

Using Genetic Programs to Evolve Satellite Algorithms

Earth Science Technology Forum

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Outline

- What led me to Genetic Programming
- NEPAC architecture
- Chlorophyll a satellite algorithms
- Genetic Programming overview
- Evolving satellite algorithms

Satellite Algorithms:

S

Box Models

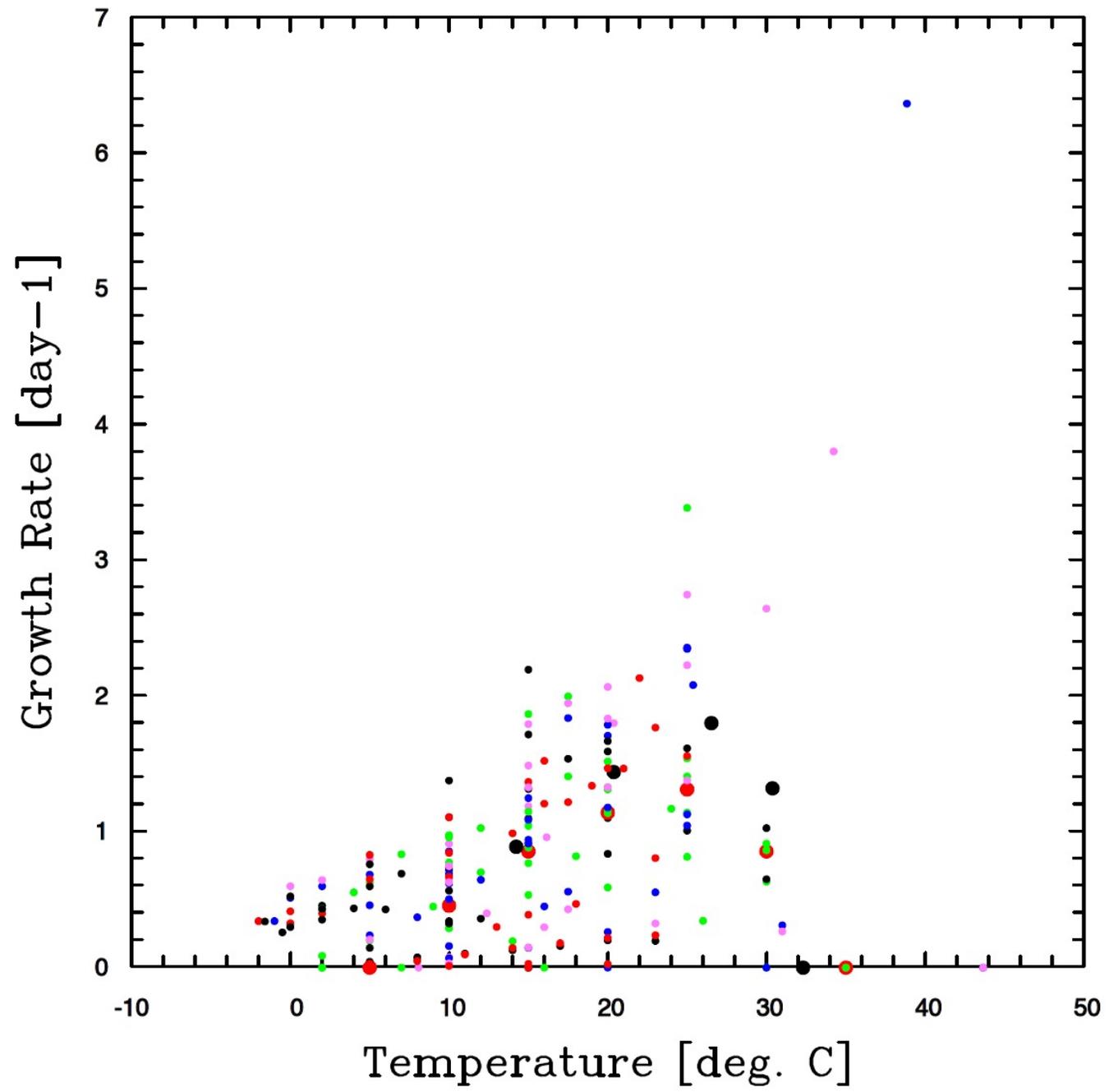
$$\frac{dB_i}{dt} = S_i$$

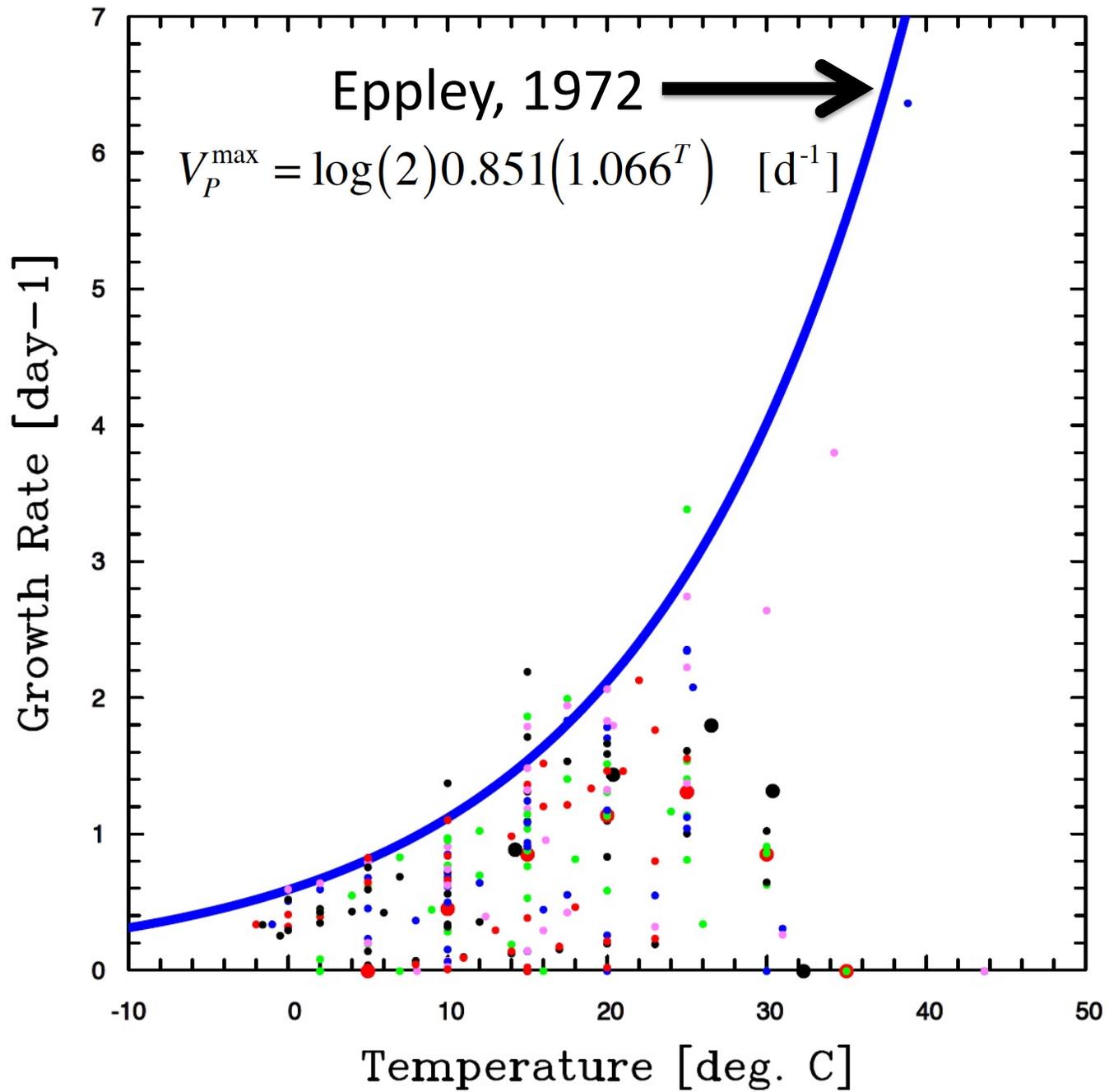
One Dimensional (Vertical) Models

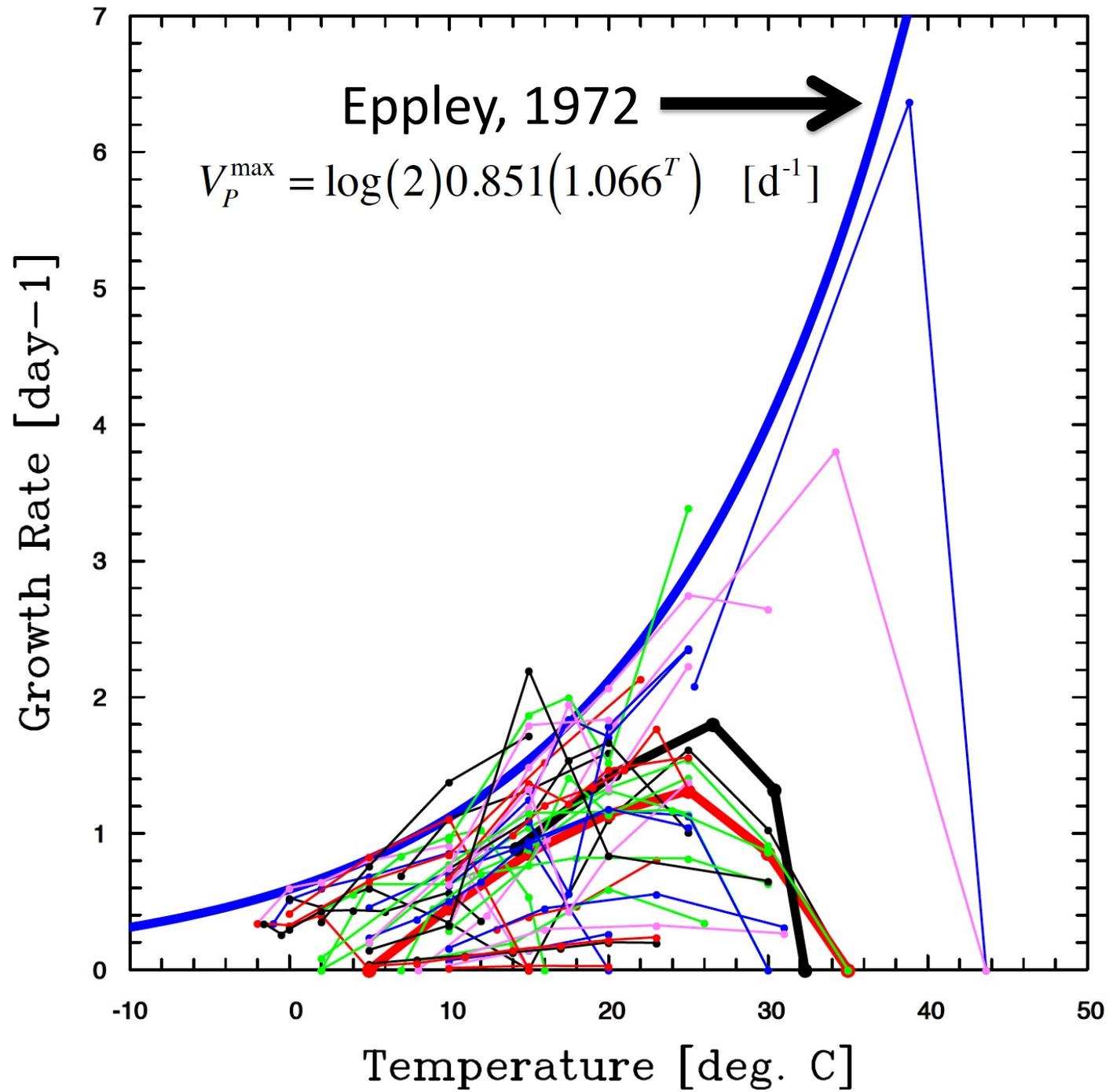
$$\frac{\partial B_i}{\partial t} + (w + w_B) \frac{\partial B_i}{\partial z} - \frac{\partial}{\partial z} K_z \frac{\partial B_i}{\partial z} = S_i$$

Three Dimensional Models

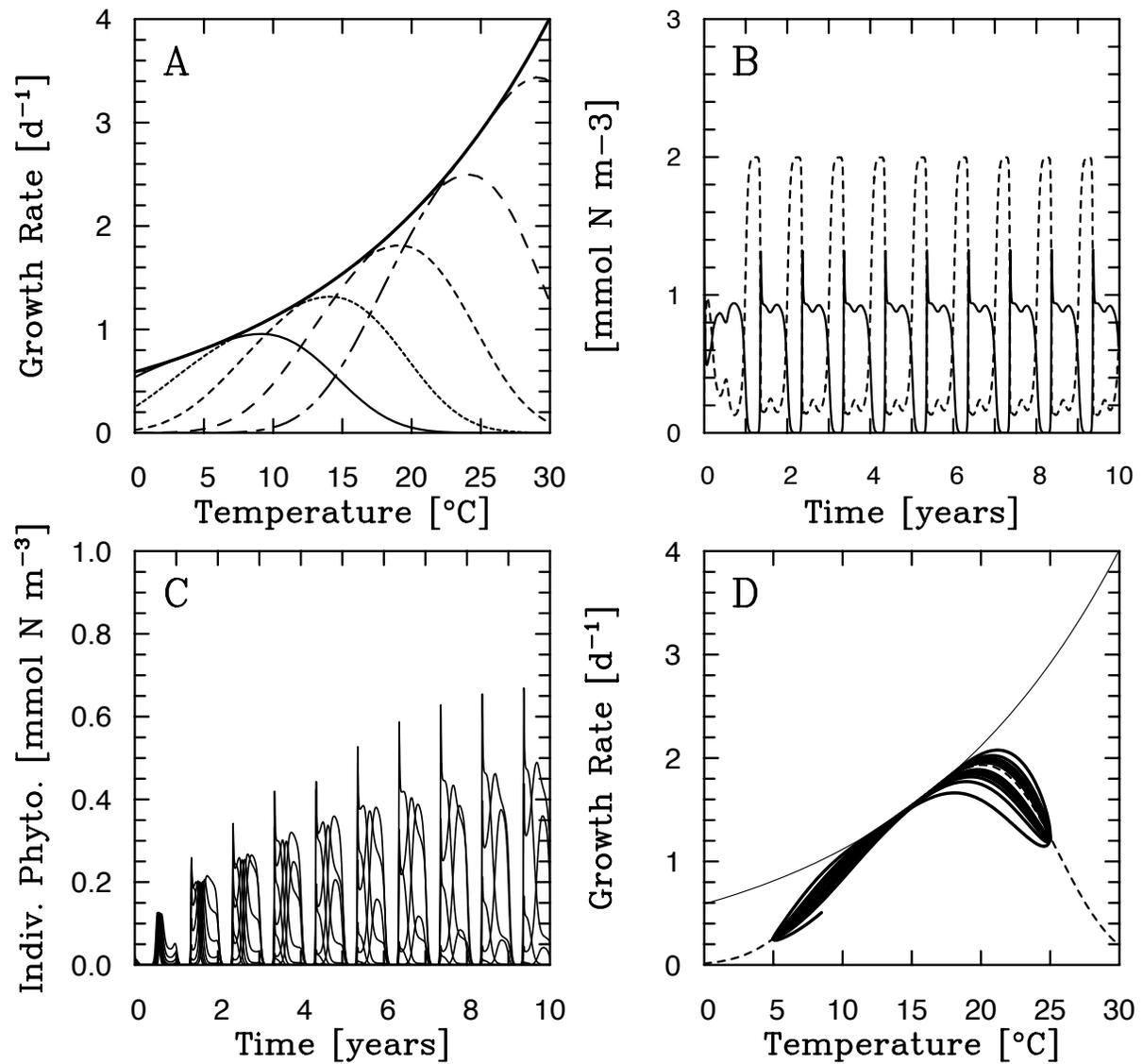
$$\frac{dB_i}{dt} + (\vec{v} + \vec{v}_B) \cdot \nabla B_i - \nabla \cdot (\vec{K} \nabla B_i) = S_i$$





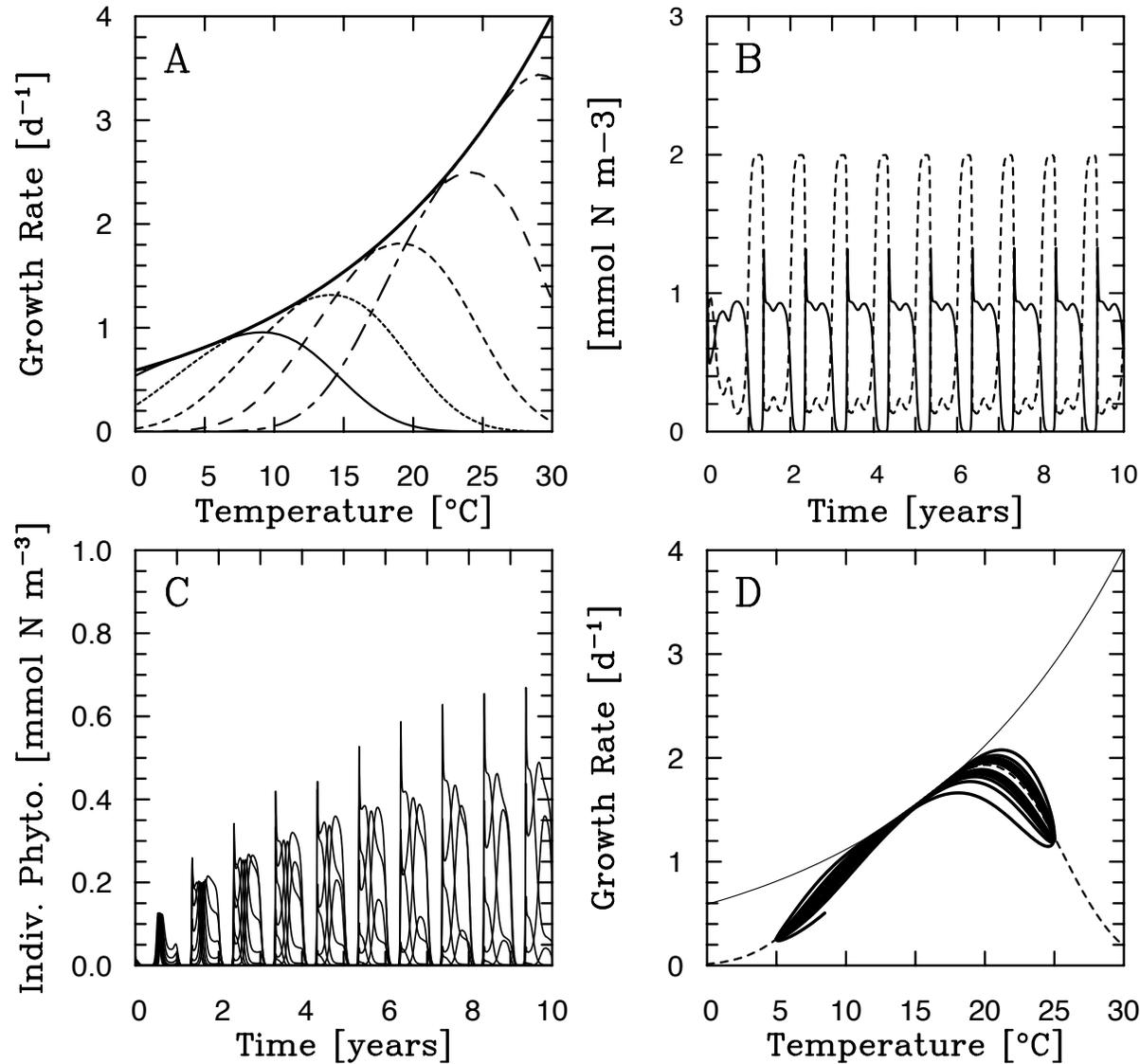


Moisan et al., 2002

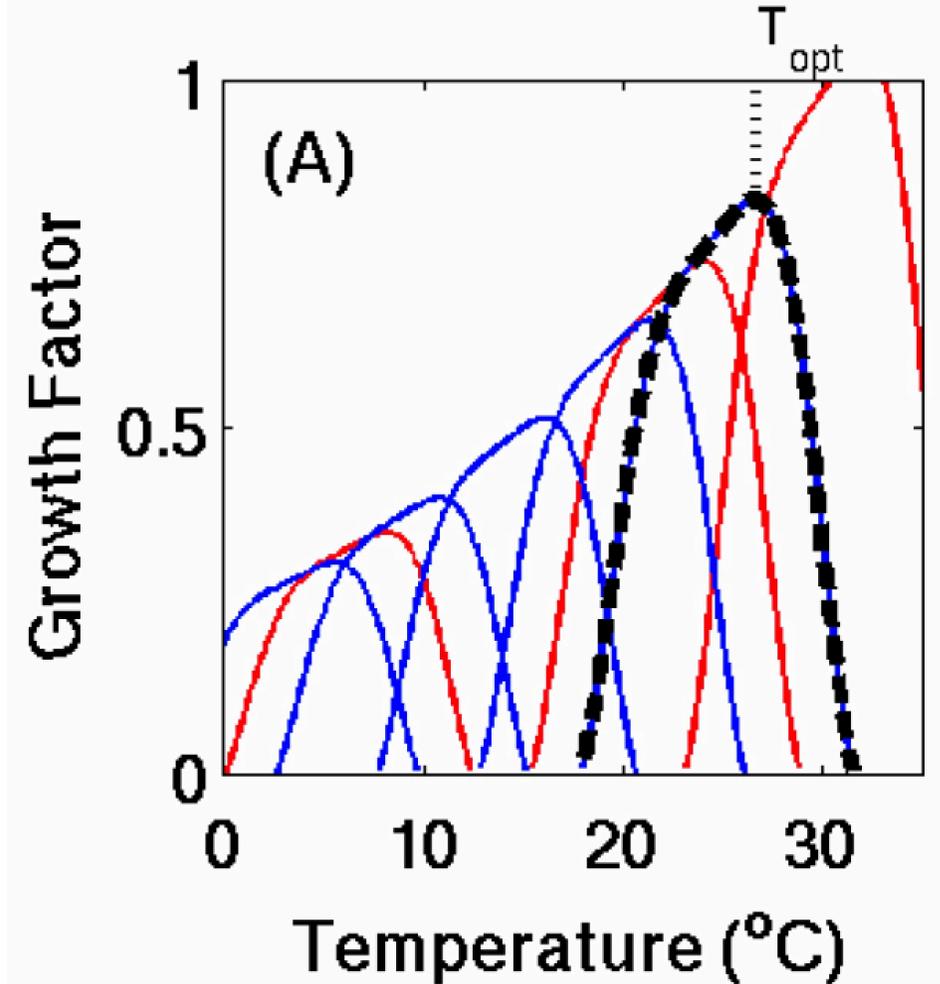


“Everything is everywhere, but the environment selects*”.

Moisan et al., 2002

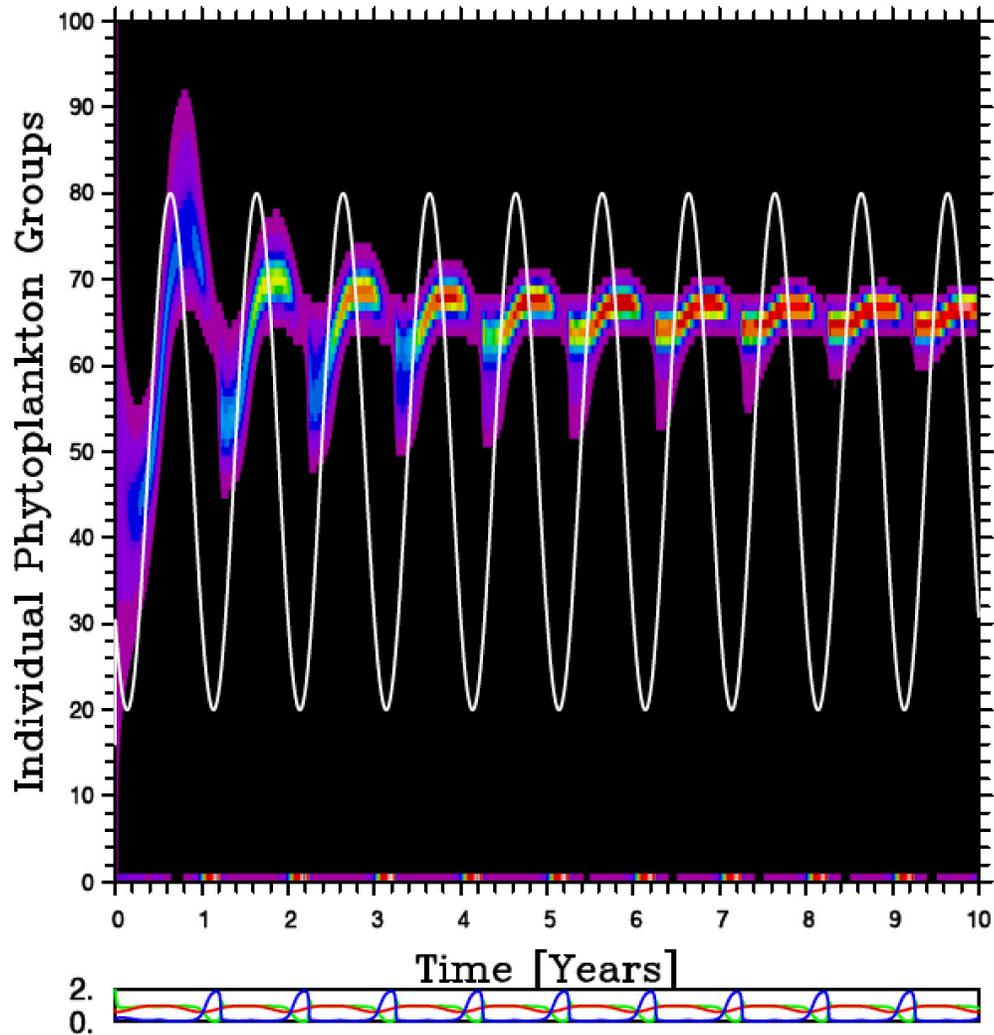


Follows et al., 2007

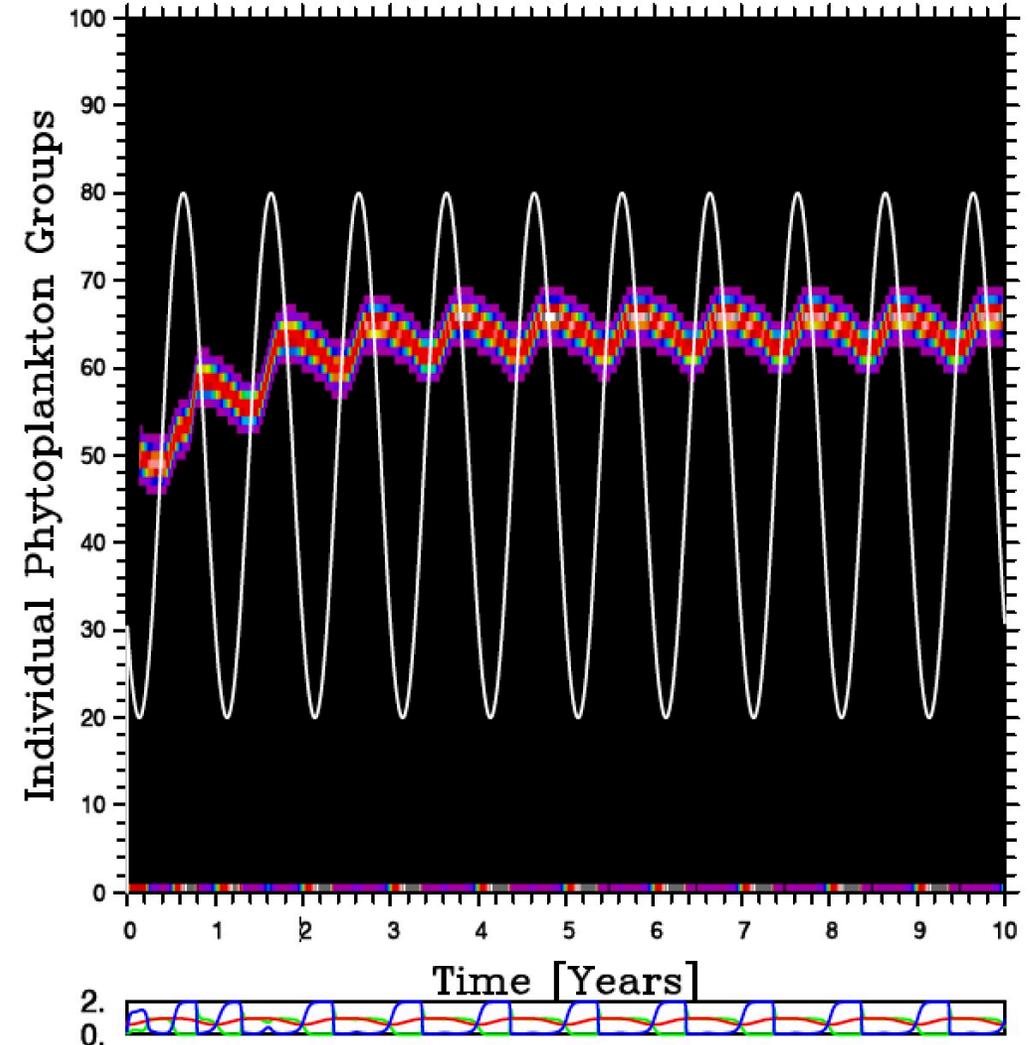


Linking parameters to traits for diversity and acclimation

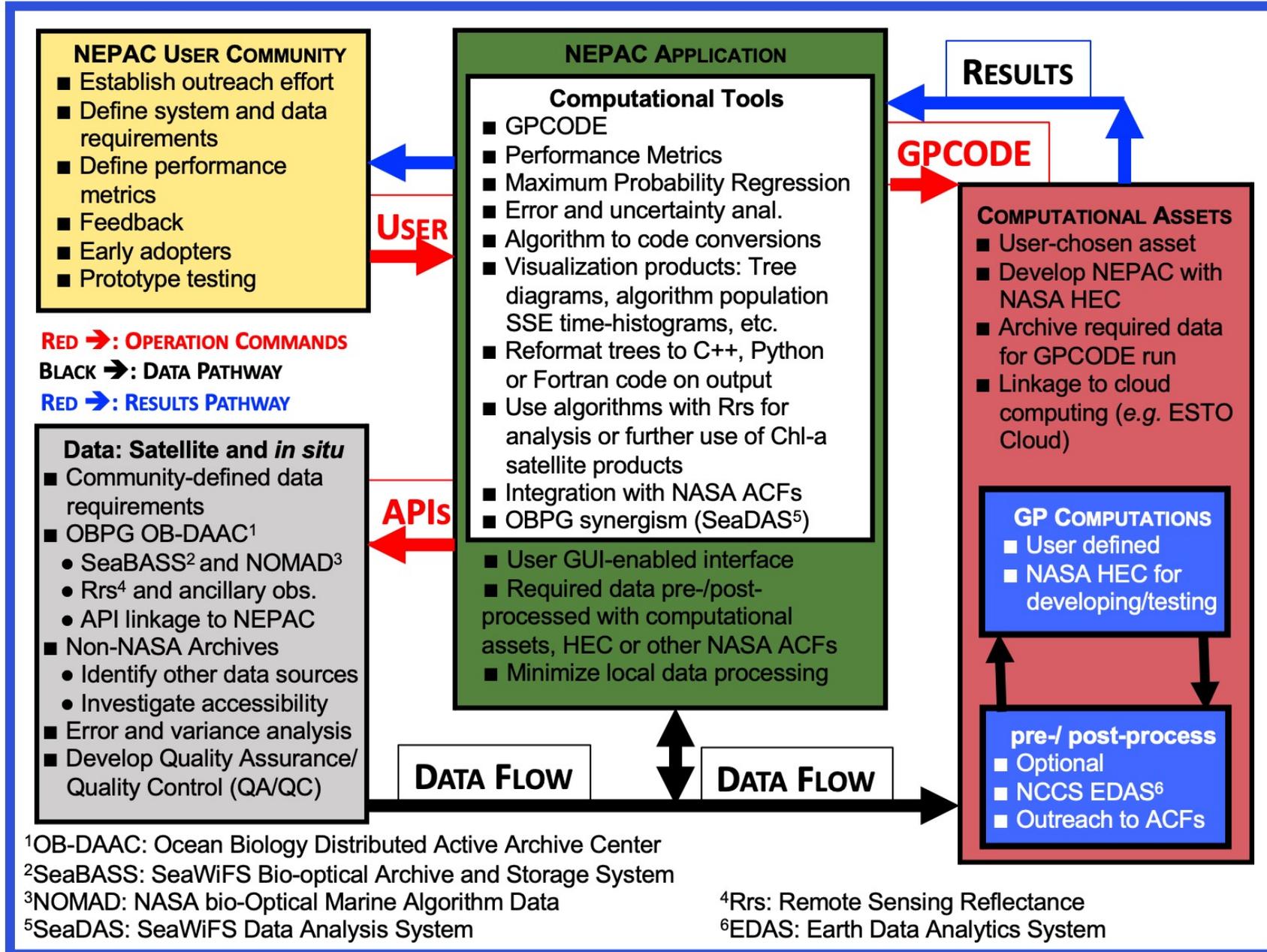
Model with 100 phytoplankton groups

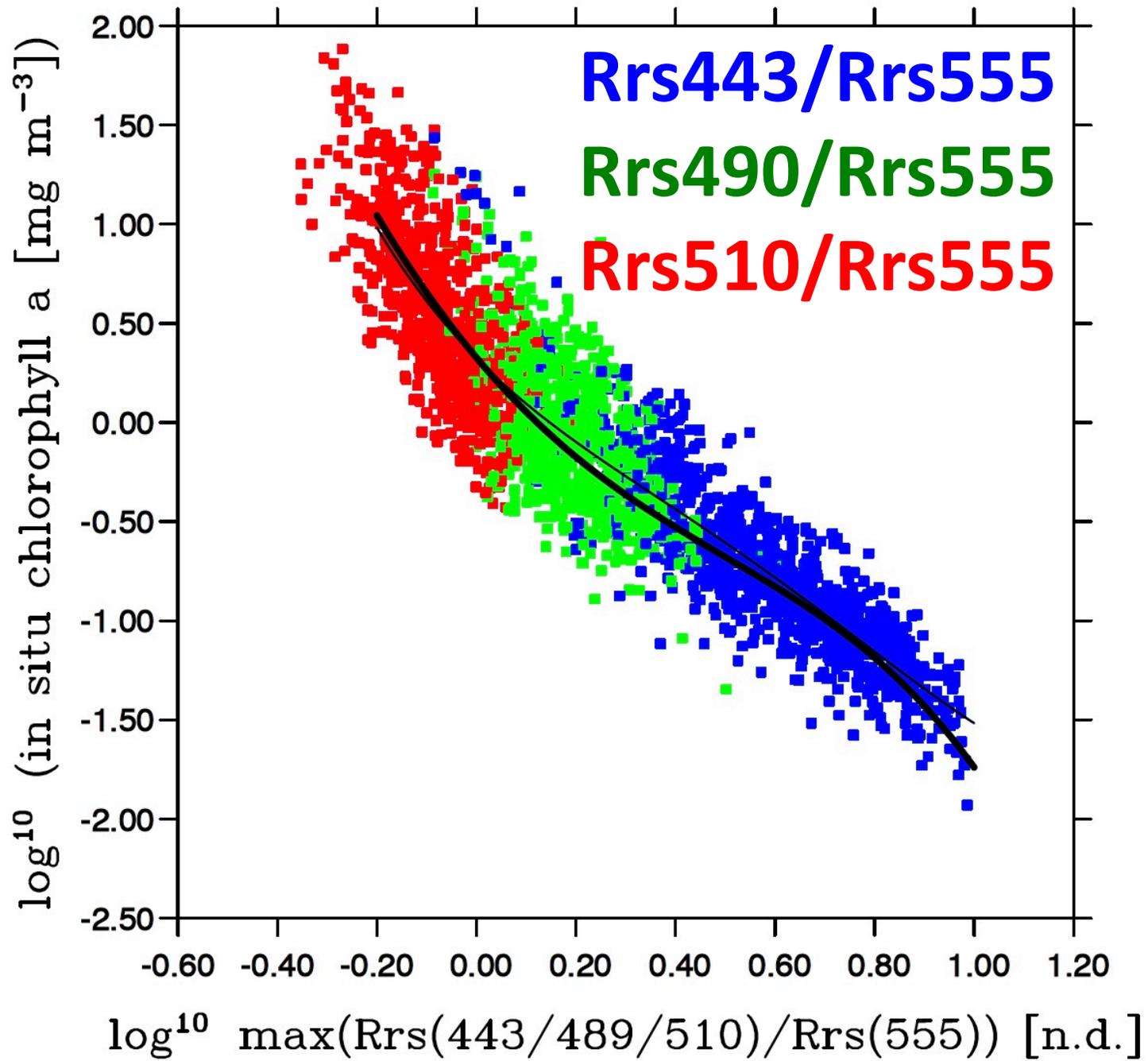


Gaussian Trait-based model (2-3 equations)



NASA Evolutionary Programming Analytic Center (NEPAC)





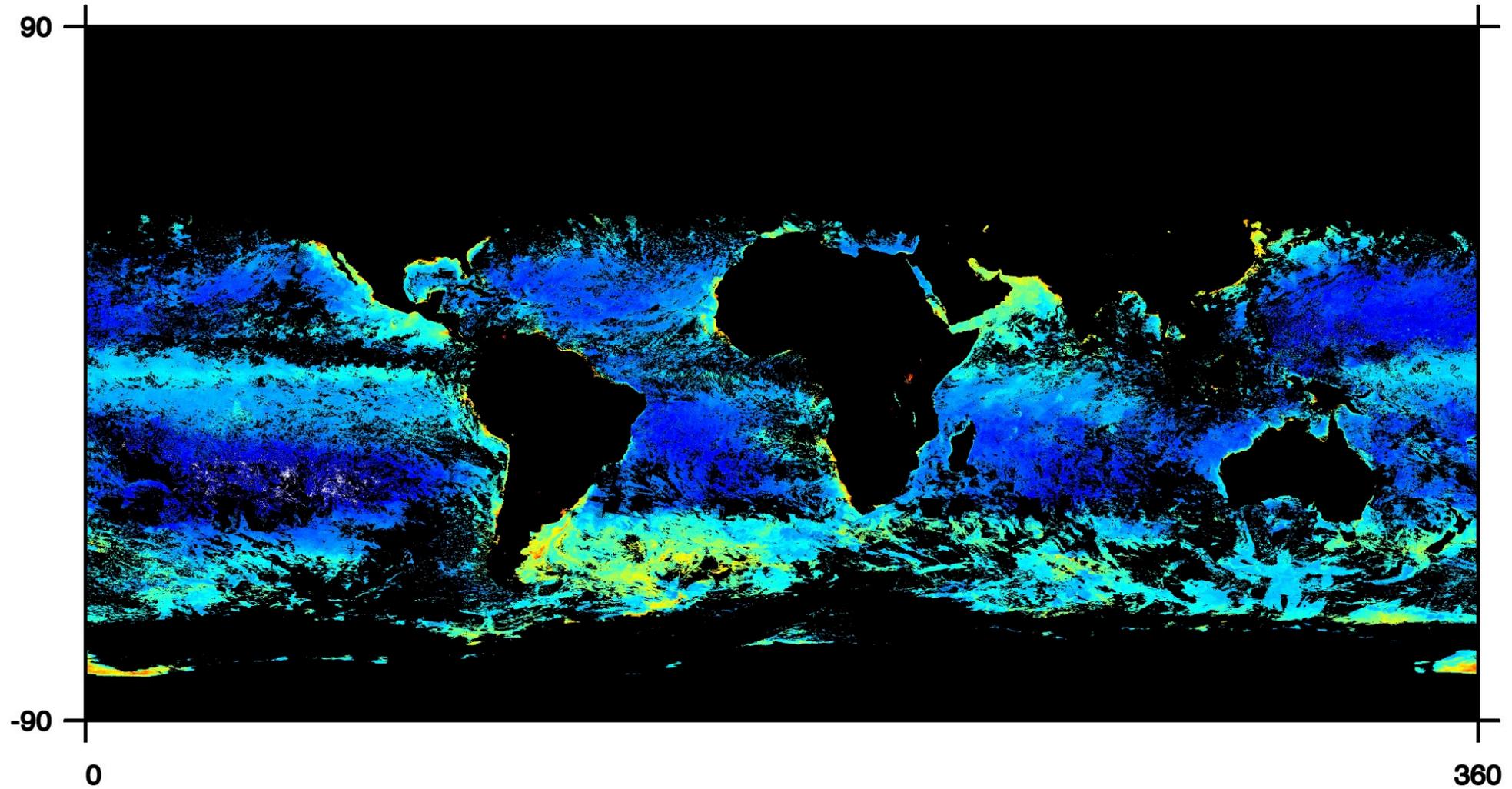
OC-4 chlorophyll a algorithm

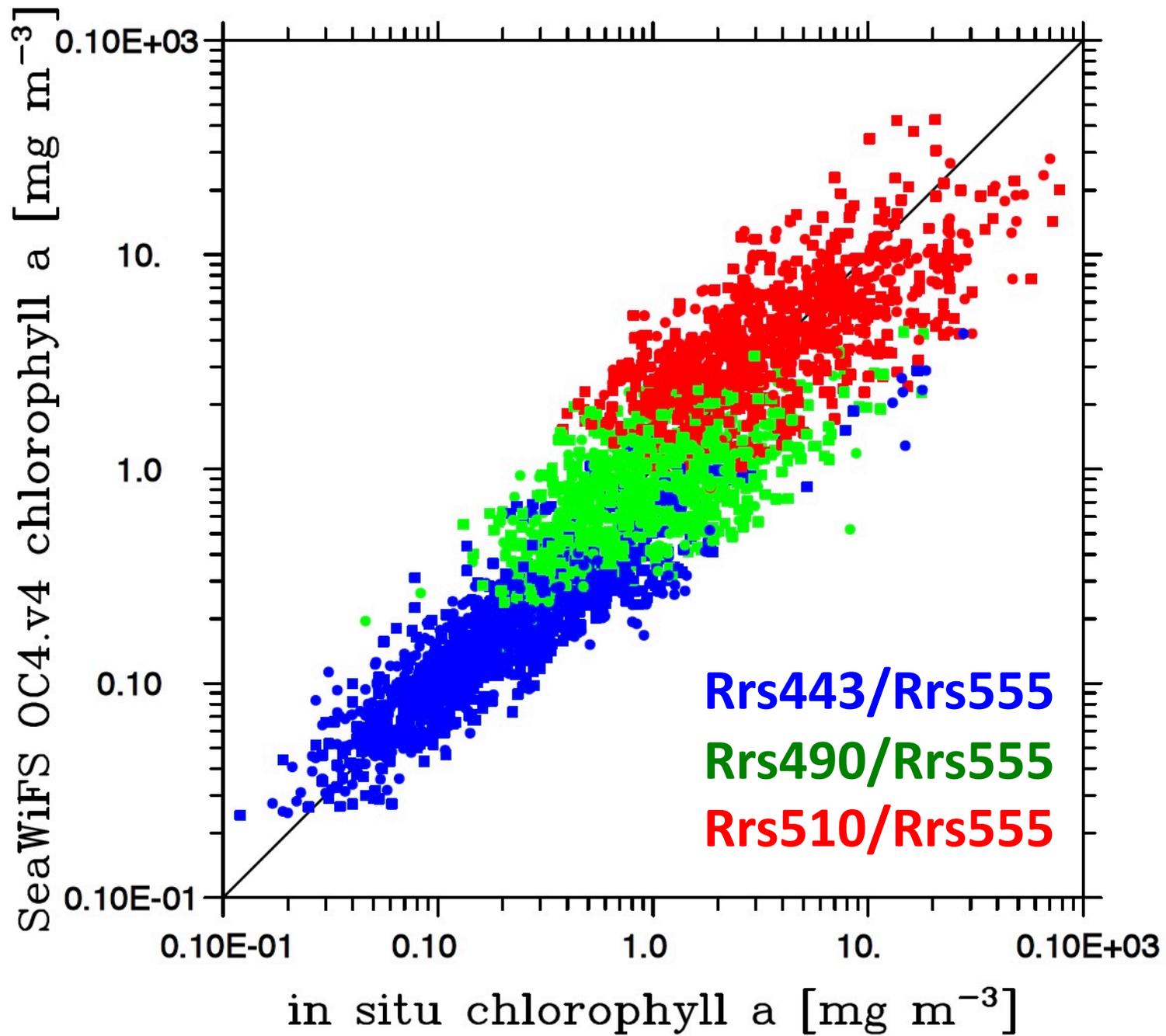
$$V = \log^{10} \left(\text{MAX} [\text{Rrs}(\lambda_{443}), \text{Rrs}(\lambda_{490}), \text{Rrs}(\lambda_{510})] / \text{Rrs}(\lambda_{555}) \right)$$

$$\log^{10}(\text{Chla}) = (a_0 + a_1 V + a_2 V^2 + a_3 V^3 + a_4 V^4)$$

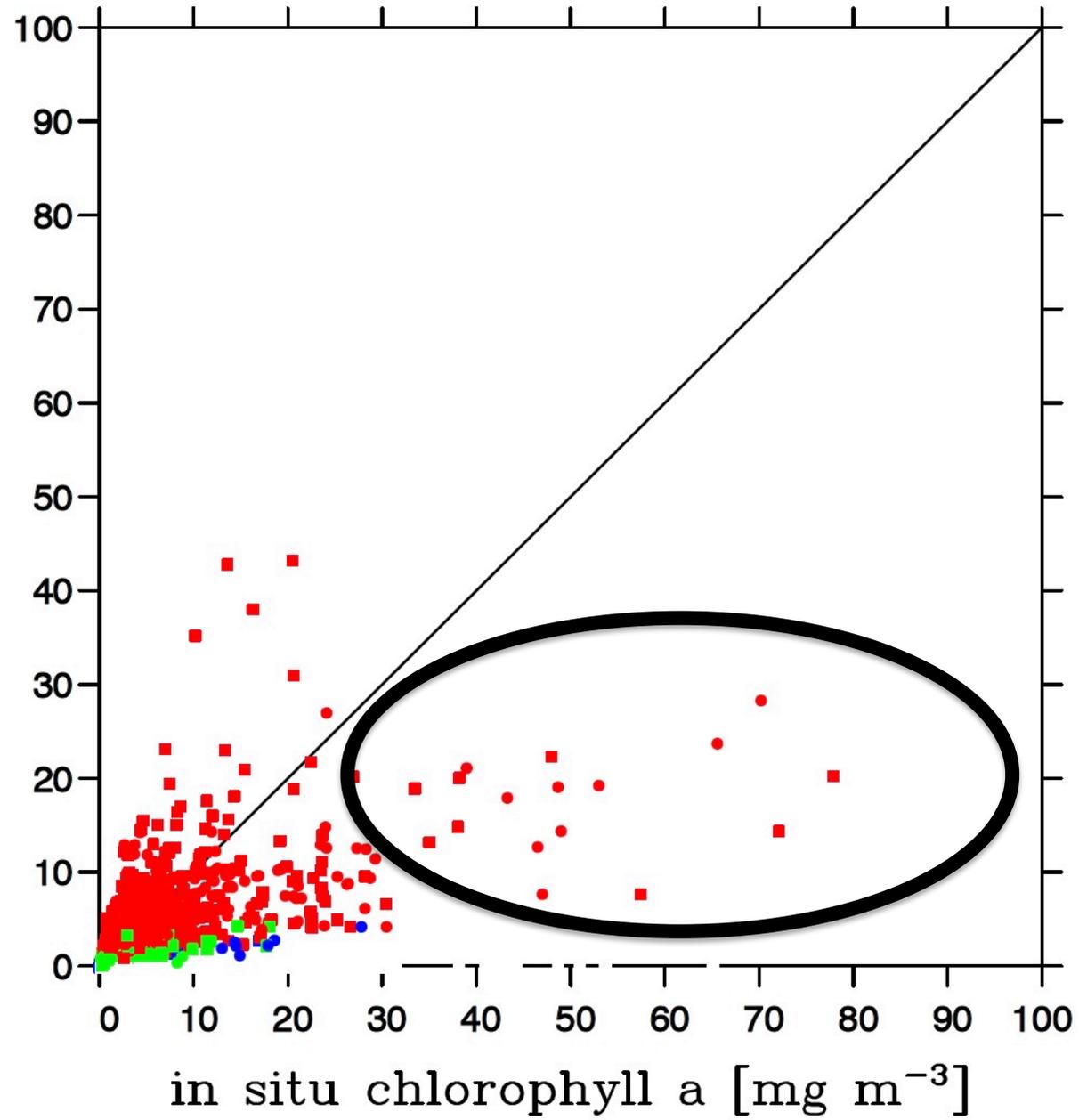
SeaWiFS OC4 Chl a [mg chla m⁻³]

0.00 [mg chla m⁻³] 63.62

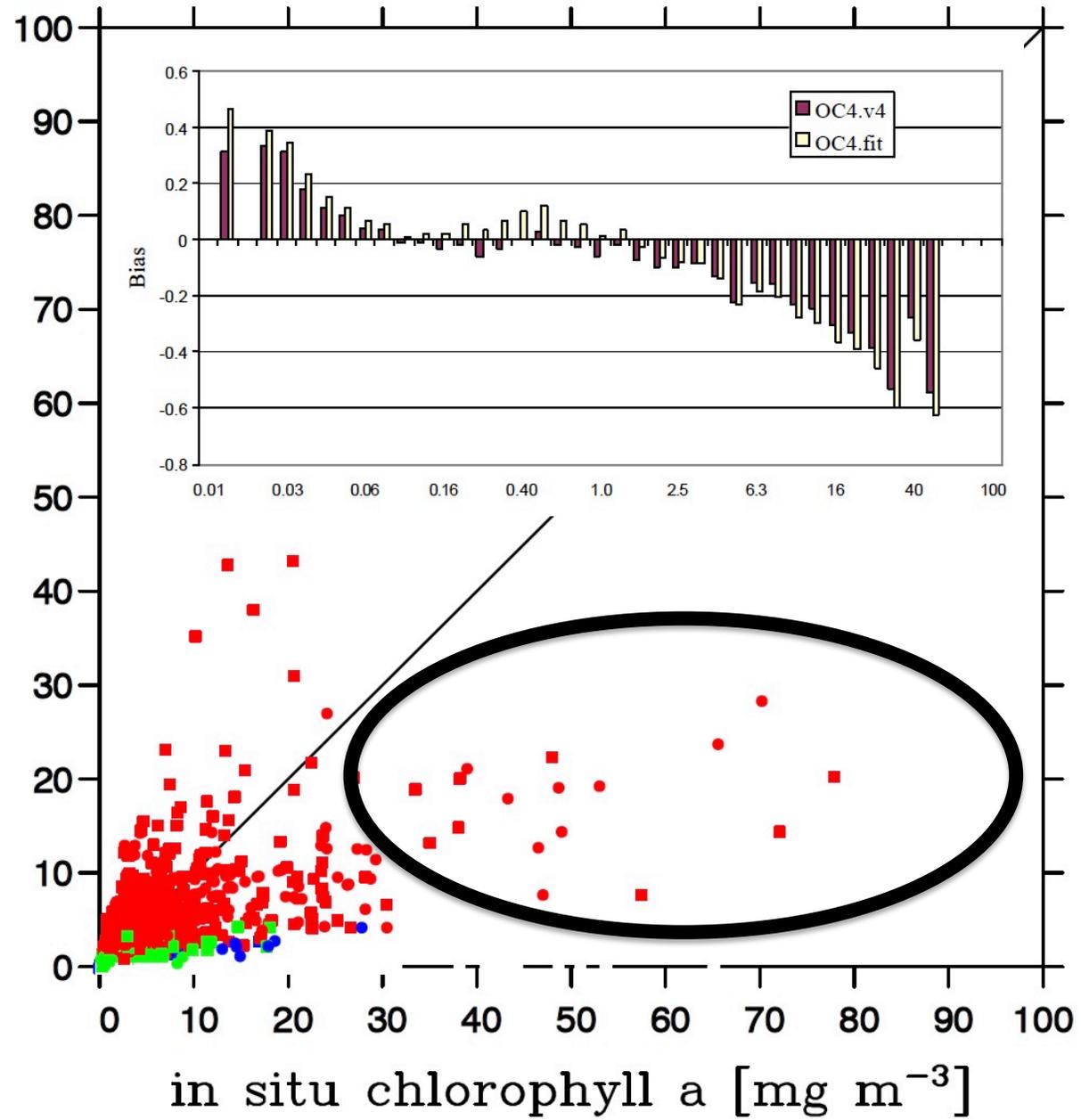




SeaWiFS OC4.v4 chlorophyll a [mg m^{-3}]

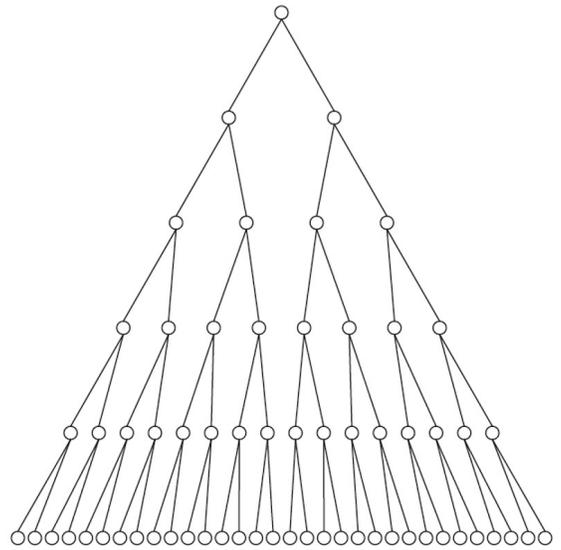


SeaWiFS OC4.v4 chlorophyll a [mg m^{-3}]



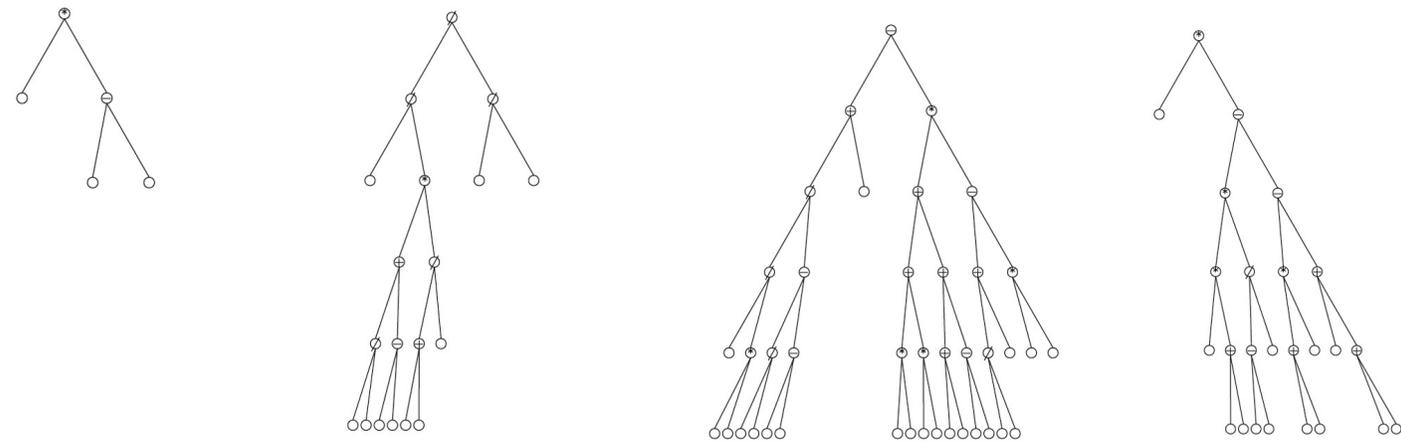
- How to create a program that can manipulate ANY equation, including ones with observations, free parameters, and functions or operators such as “+”, “-”, “*”, “/”, and even items such as Booleans “IF/THEN”, etc?

Binary Tree Architecture



Empty and Full Binary Tree Map

Example of randomly generated initial tree structures



Use of Fortran-90 'Defined Type'

Operator:

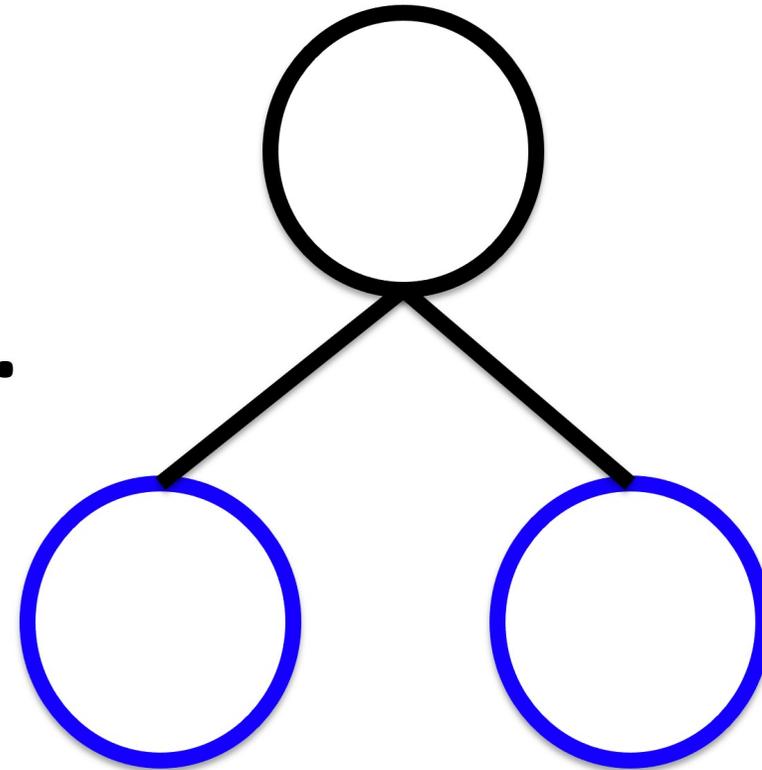
[+, -, *, /, Boolean, etc]

Input nodes can vary.

Input Nodes:

[Variable, parameter,

New defined type]



Calculation starts from top of tree

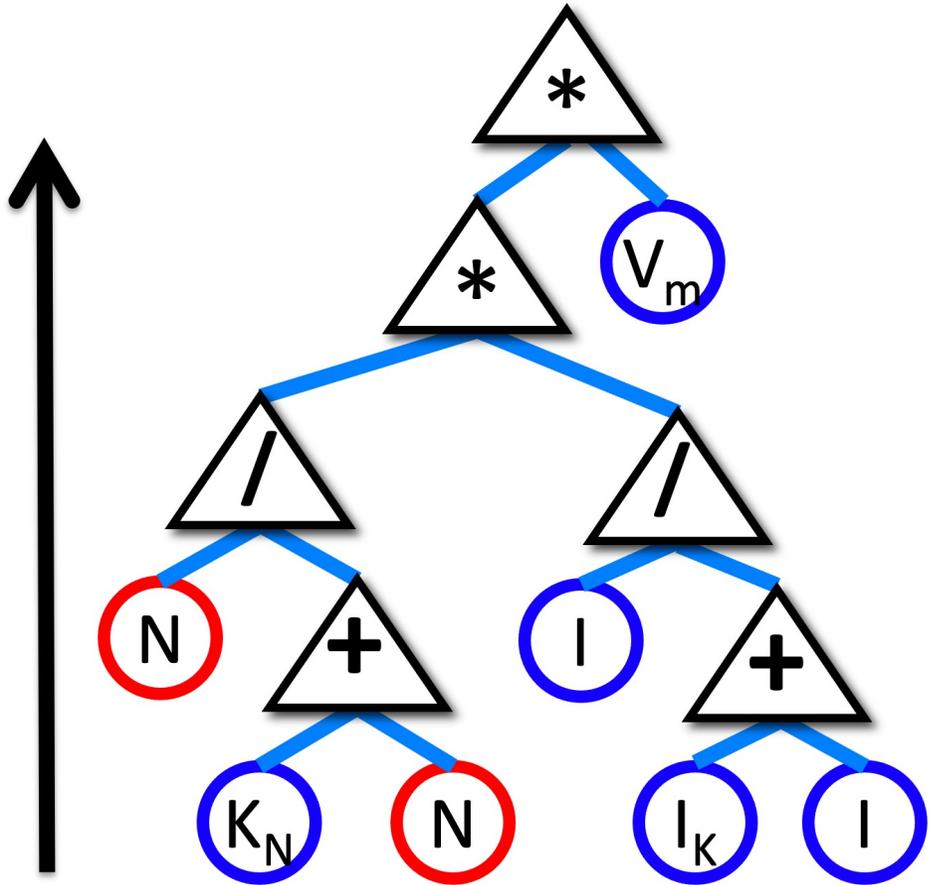
Performance improved by orders of magnitude!

$$\text{Growth} = V_m \left[\frac{N}{K_N + N} \right] \left[\frac{I}{I_K + I} \right]$$

LISP prefix representation:

(* V_m (* (/ N (+ (K_N N))) (/ I (+ (I_K I))))))

Calculation of Tree value moves from the bottom up.

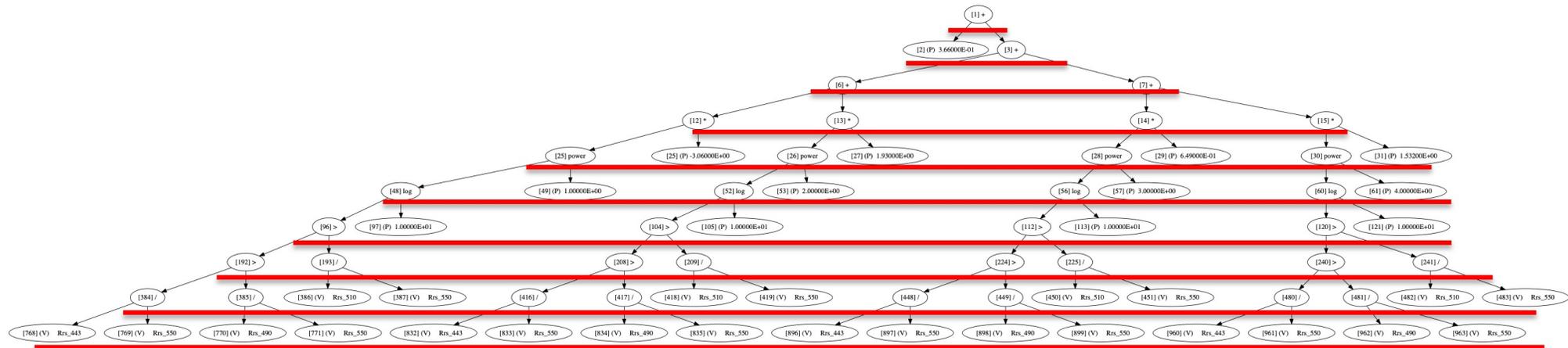


OC-4 chlorophyll a algorithm

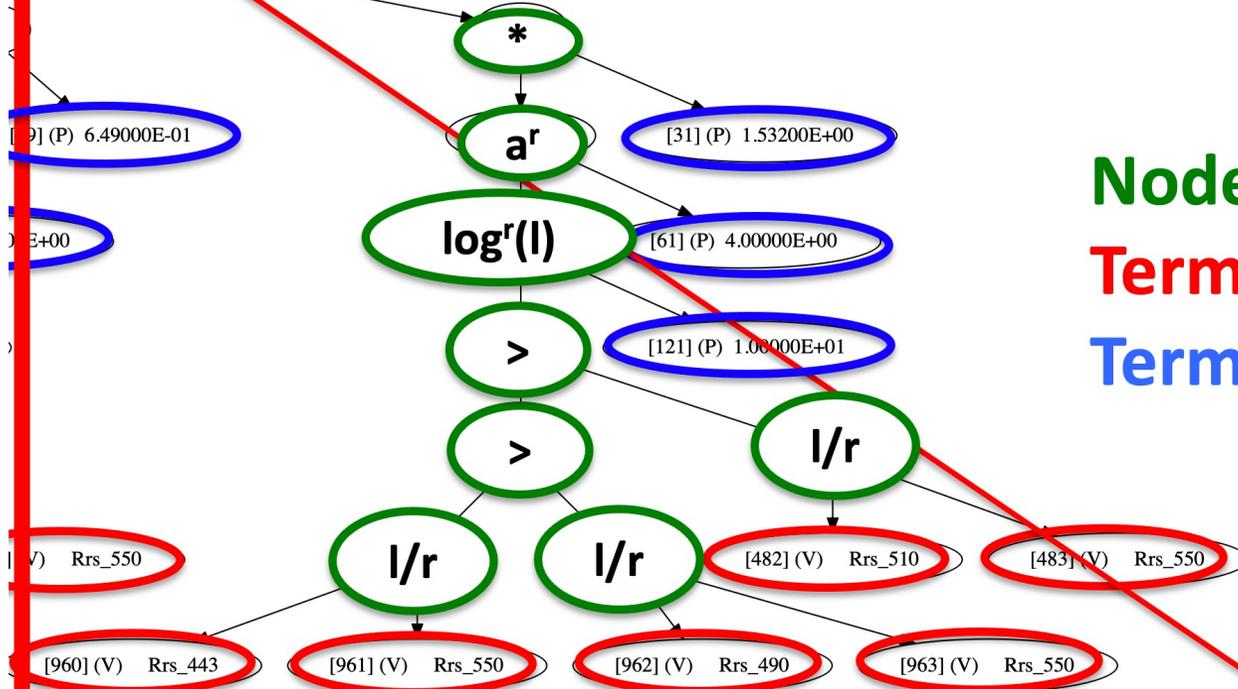
$$V = \log^{10}(\text{MAX}[\text{Rrs}(\lambda_{443}), \text{Rrs}(\lambda_{490}), \text{Rrs}(\lambda_{510})] / \text{Rrs}(\lambda_{555}))$$

$$\log^{10}(\text{Chla}) = (a_0 + a_1 V + a_2 V^2 + a_3 V^3 + a_4 V^4)$$

OC-4 Algorithm Tree \rightarrow 10 levels



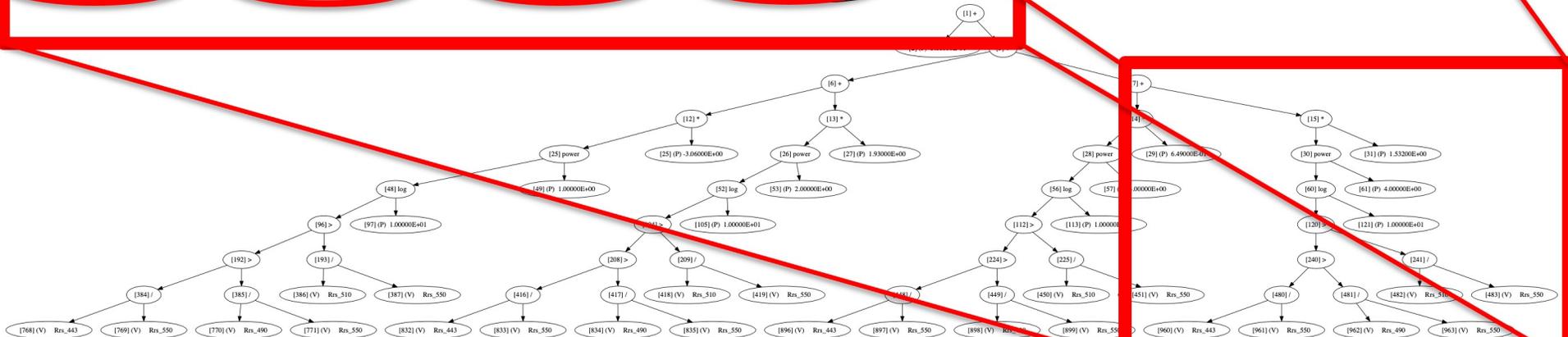
OC4 Algorithm Tree



Node Functions

Terminal Variables (Rrs)

Terminal Parameters



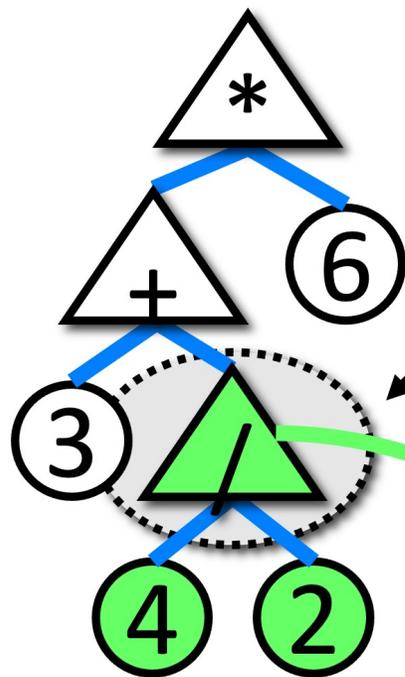
Genetic Programming Overview

- 1) Generate initial random population of equations/models
- 2) Calculate fitness of all individual equations/model
 - Strip out all model/equations parameters
 - Carry out Genetic Algorithm Optimization
 - Finish off with localized LMDIF Optimization
- 3) Randomly select, based on fitness, for:
 - Asexual reproduction
 - Sexual reproduction (i.e. Tournament Selection)
 - Mutations
- 4) Calculate fitness of new individuals (same as 2 above)
- 5) Test for completion criteria (low SSE value met?)
- 6) No: go to 3; Yes: end program

Cross methodology

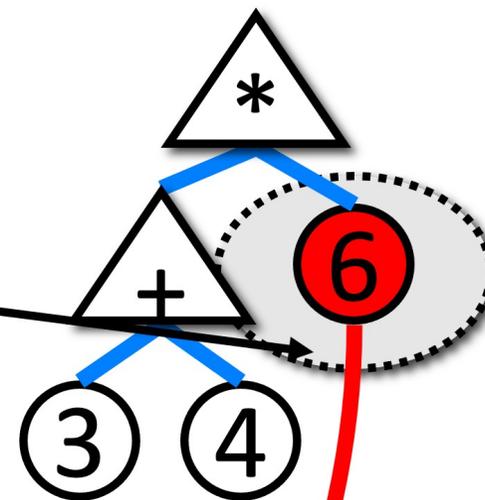
Before Cross

Parent #1



Parent #2

Randomly Chosen Nodes or Terminals

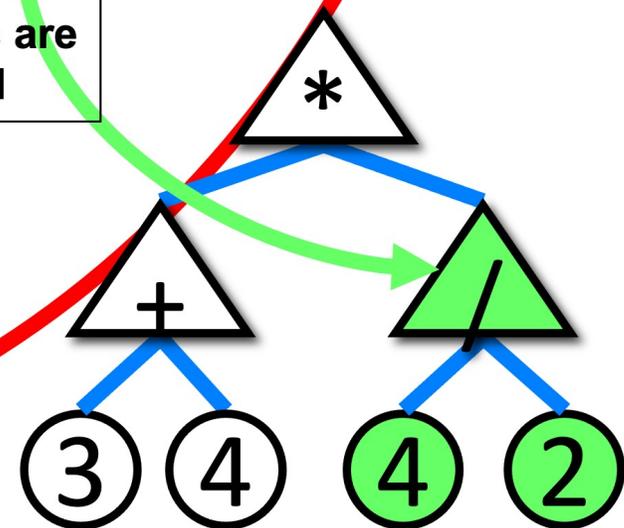
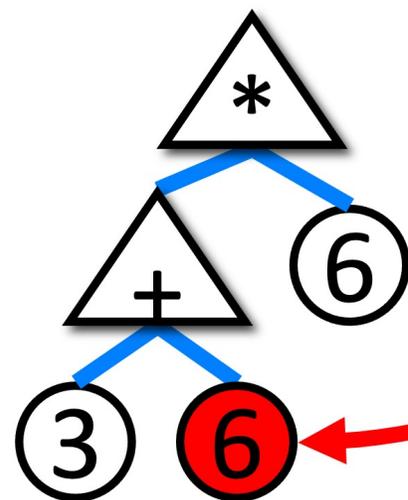


Child #1

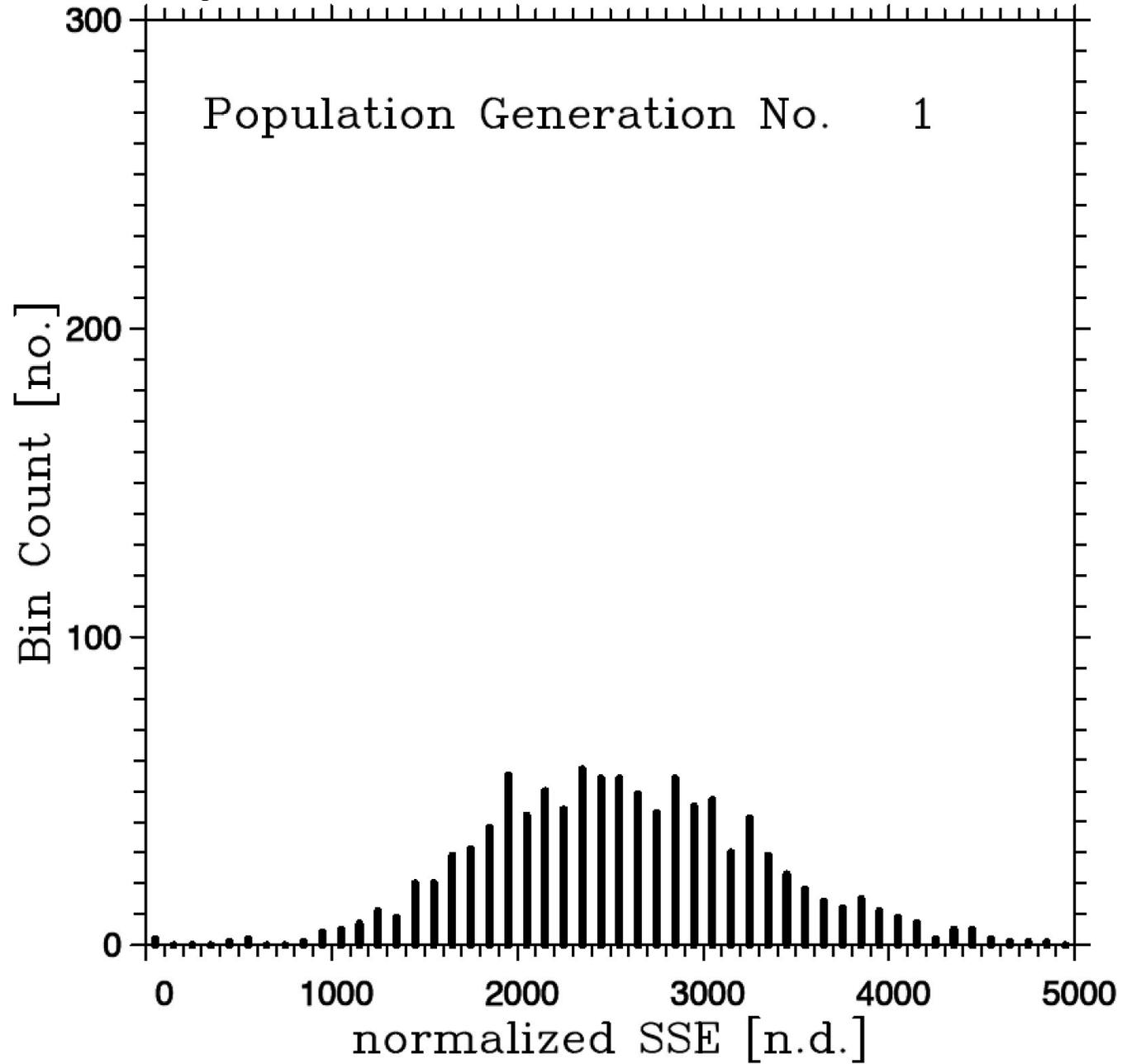
Related Subtrees are Swapped

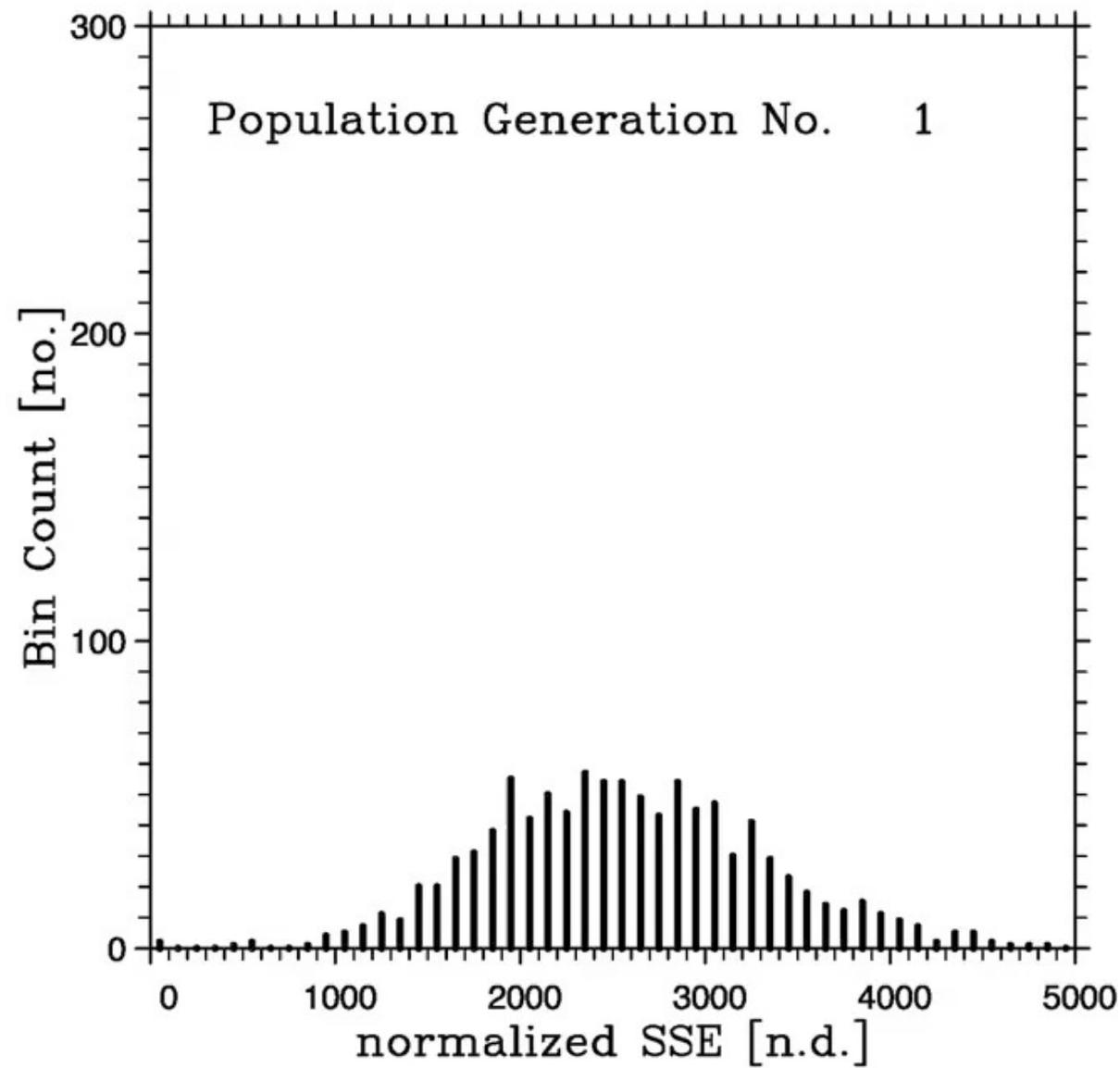
Child #2

After Cross

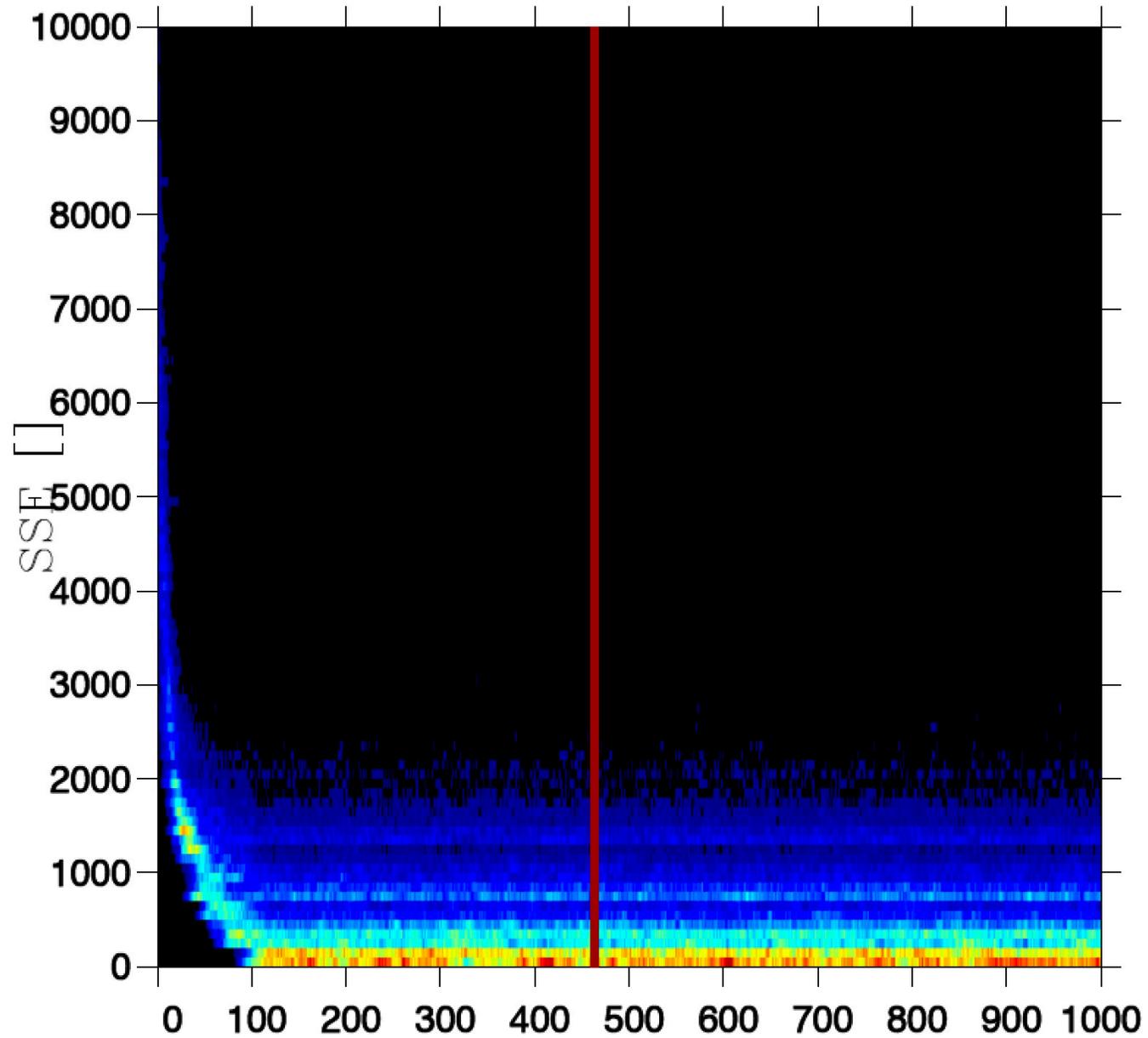


Simple Lotka-Volterra Model

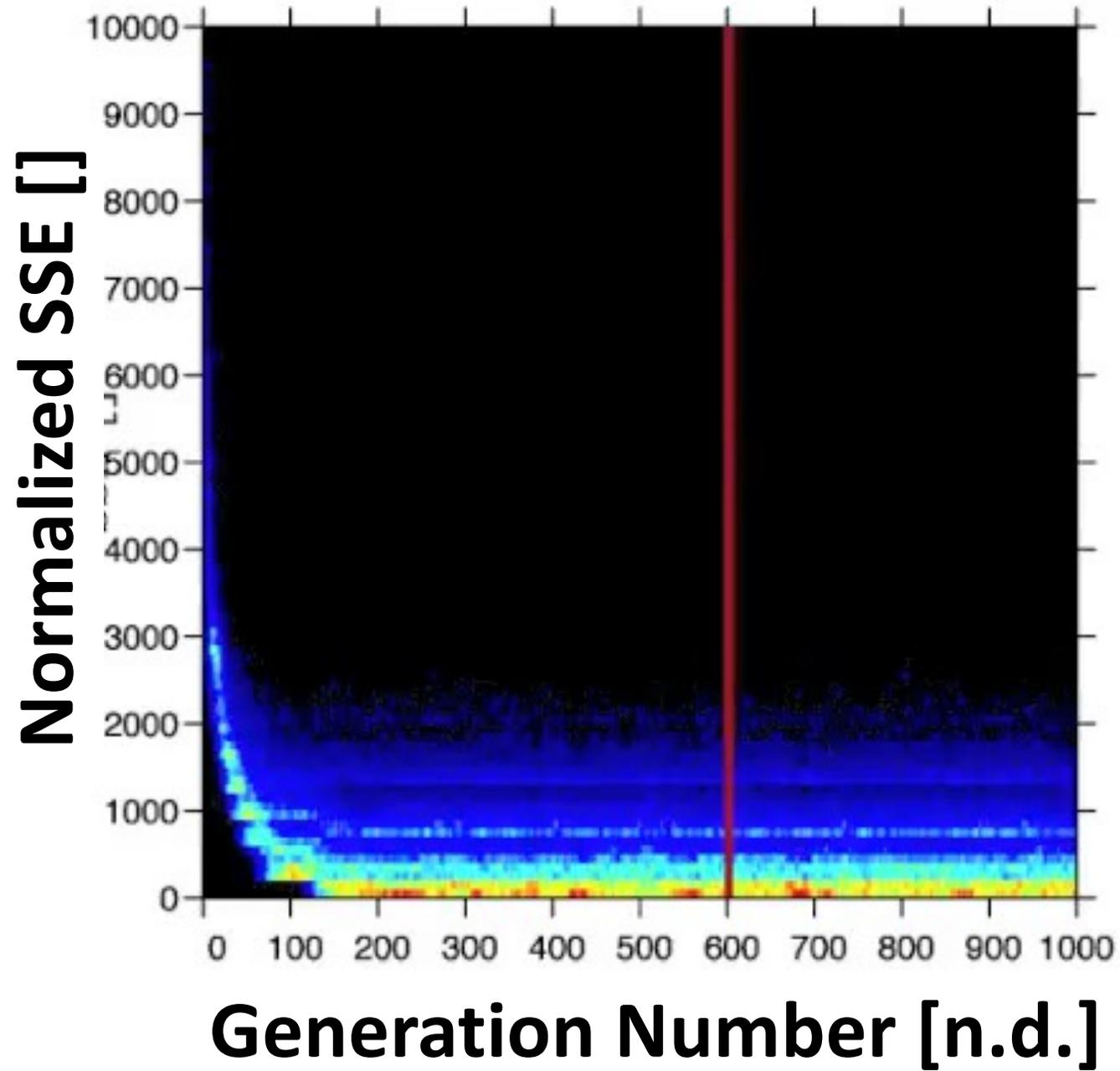




Normalized SSE []



Generation Number [n.d.]



GPCODE Chla algorithms using IOCCG data for training

Case 12: GP_Tree_All_SSE

n_GP_Generations=304

n_GP_Individuals=1000

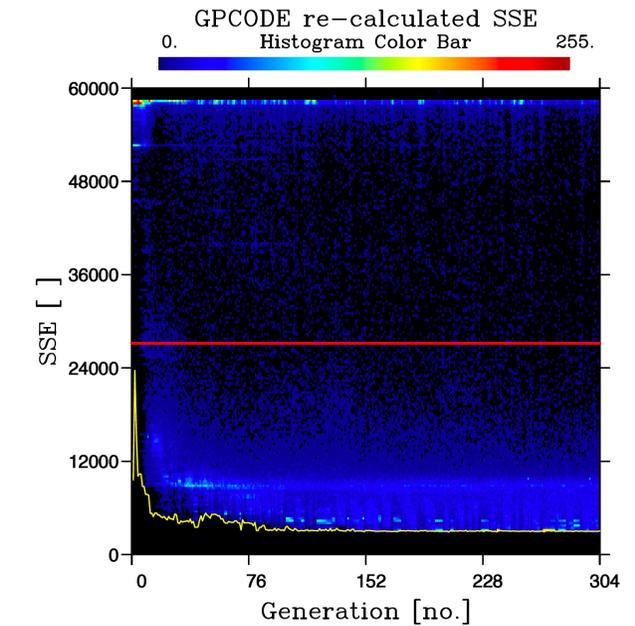
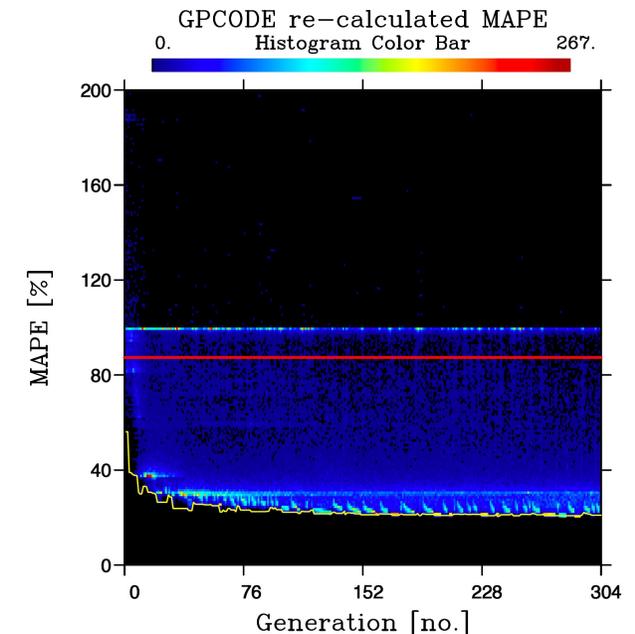
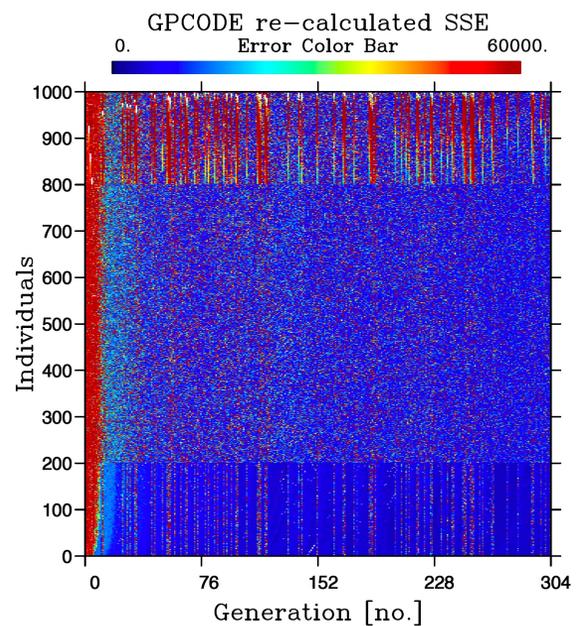
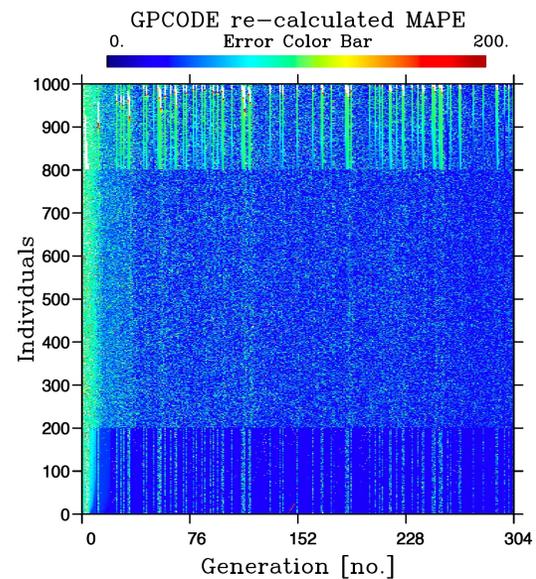
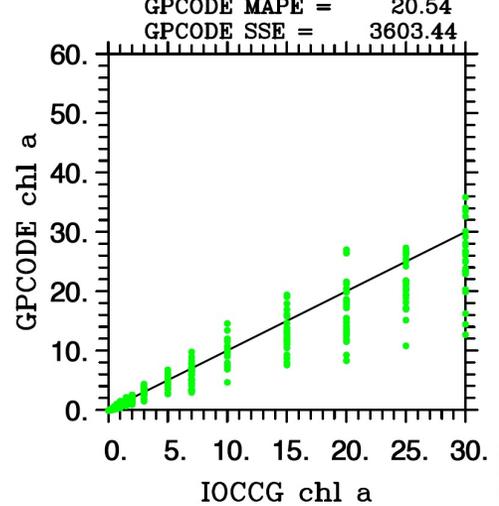
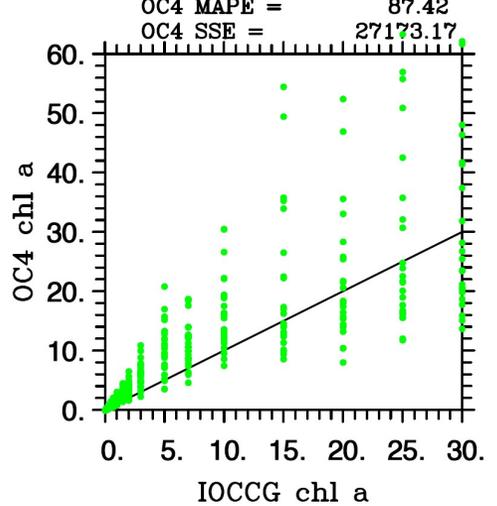
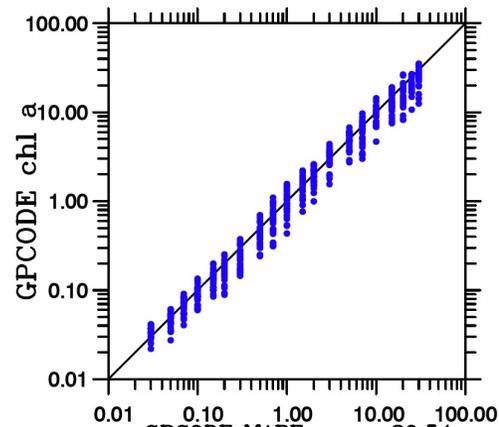
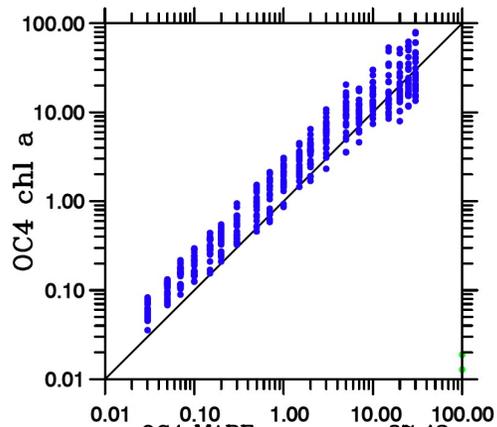
MAPE metric

No Bootstrapping (BS)

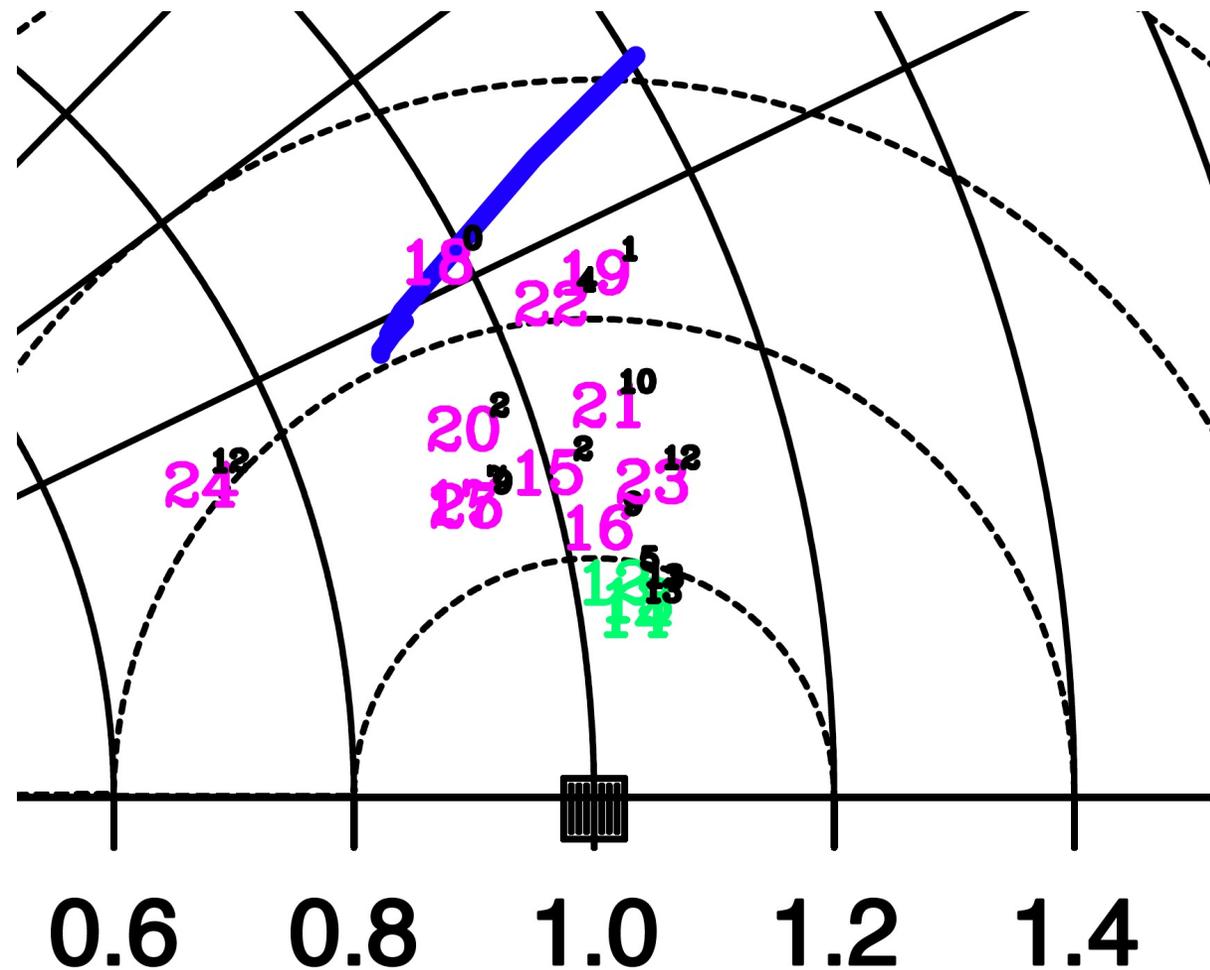
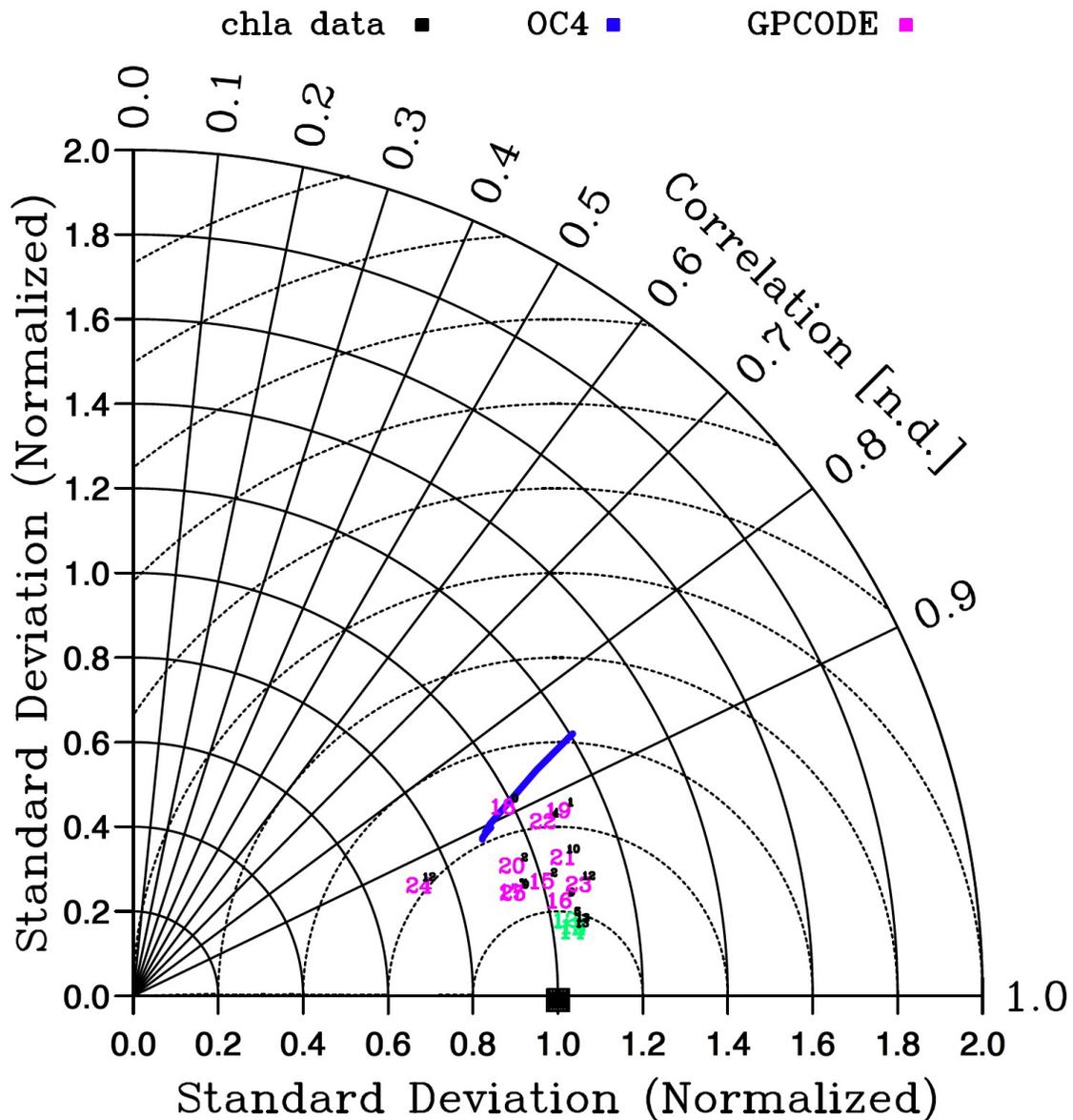
of parameters in best tree: 9

Best SSE Tree (gen. = 238 , ind. = 1)

Best MAPE Tree (gen. = 206 , ind. = 614)

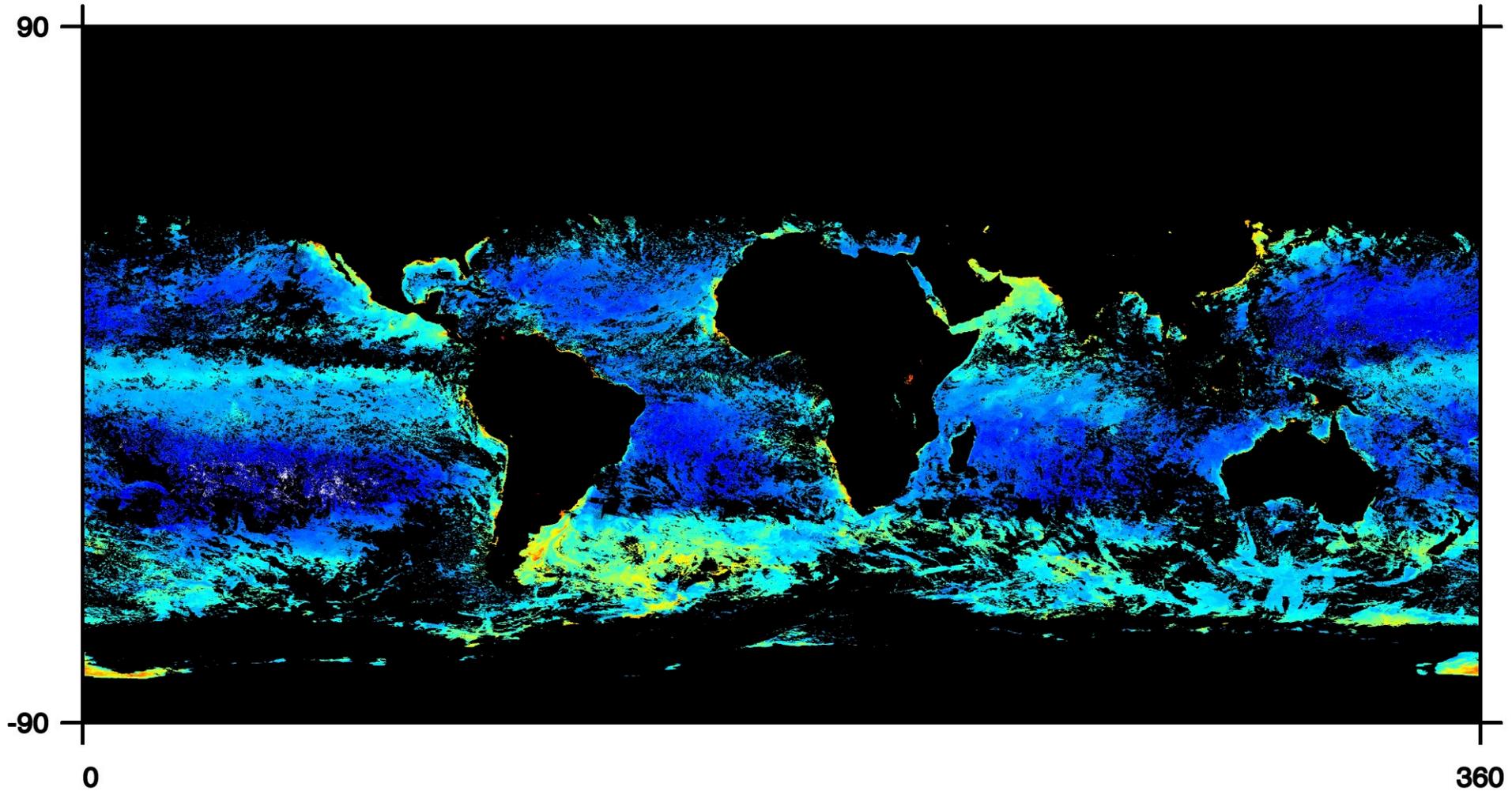


GPCODE chla algorithms using IOCCG data



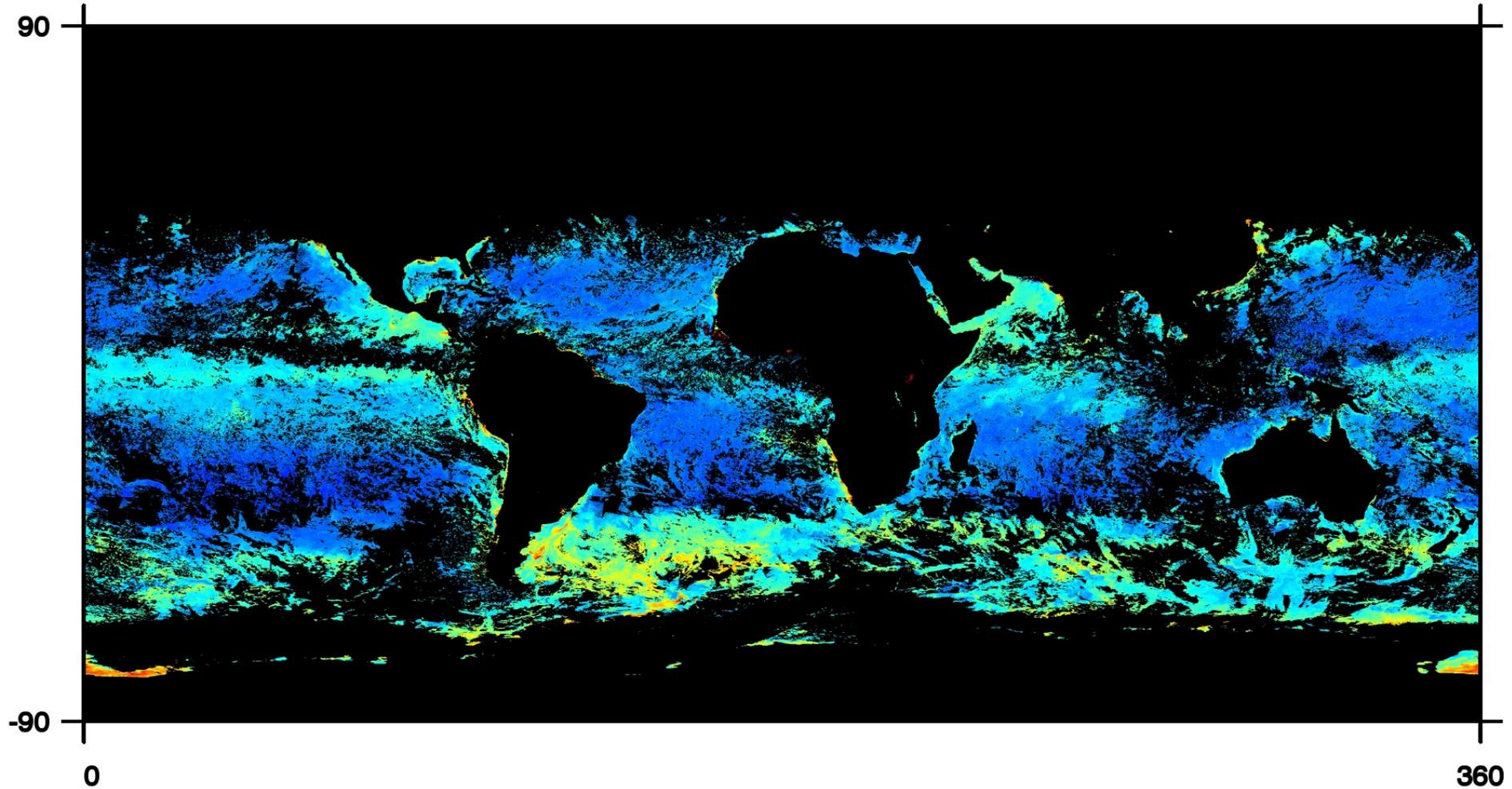
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0.00 [mg chla m⁻³] 63.62



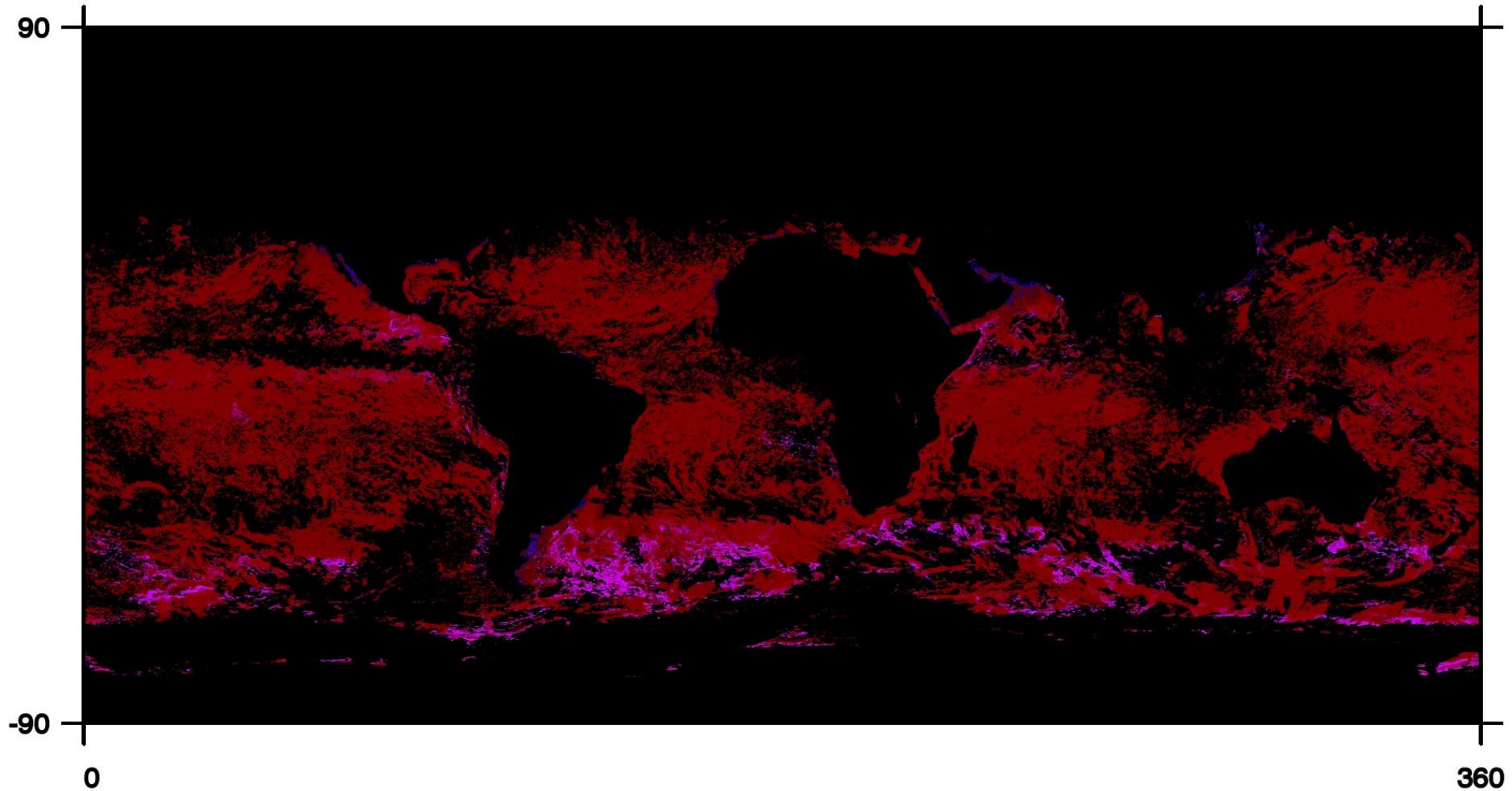
SeaWiFS GPCODE Chl a [mg chla m⁻³]

0.00 [mg chla m⁻³] 74.99



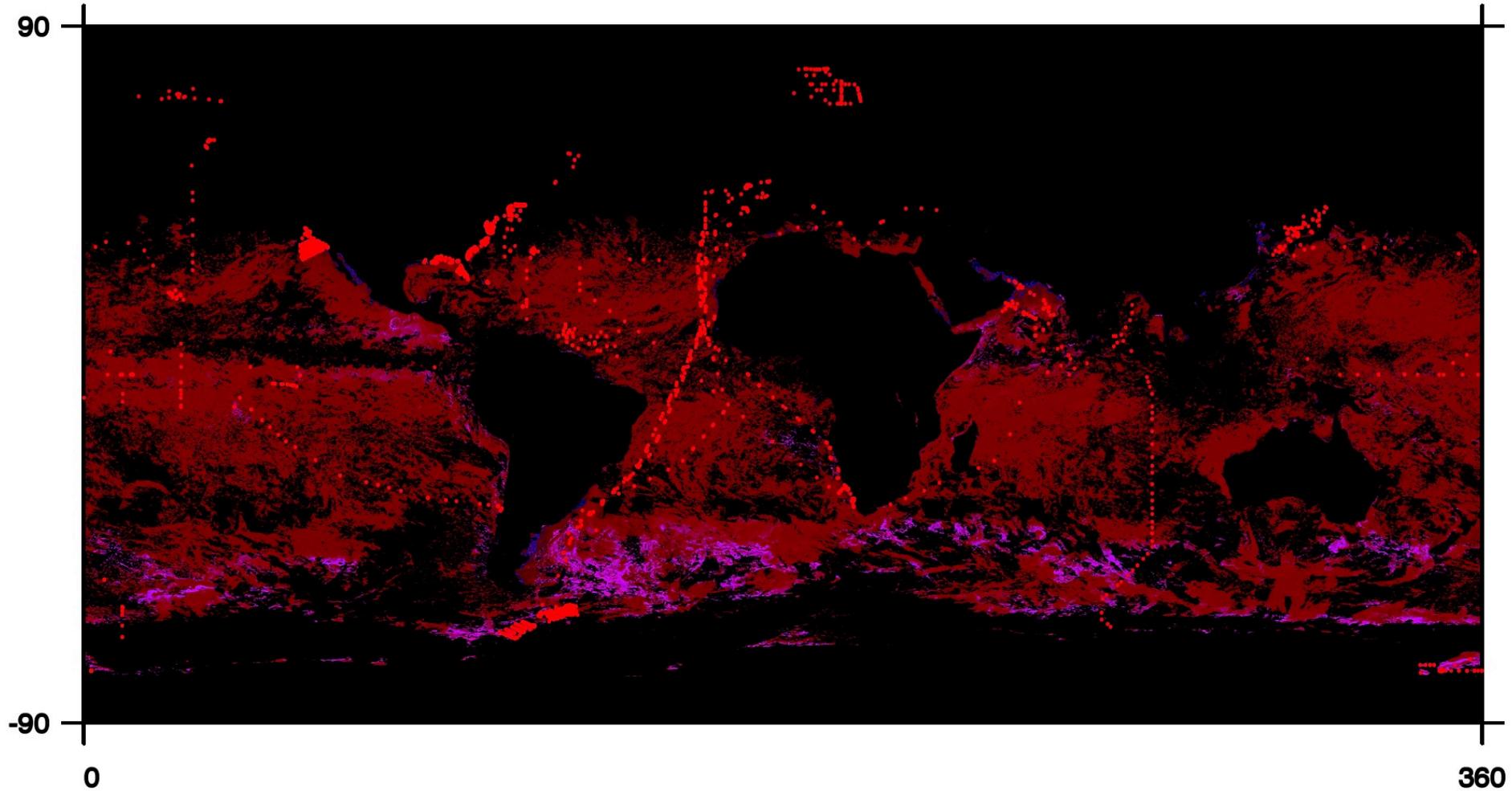
SeaWiFS GPCODE-OC4 Chl a [diff]

-1.00 [mg chla m⁻³] 1.00



SeaWiFS GPCODE-OC4 Chl a [diff]

-1.00 [mg chla m⁻³] 1.00



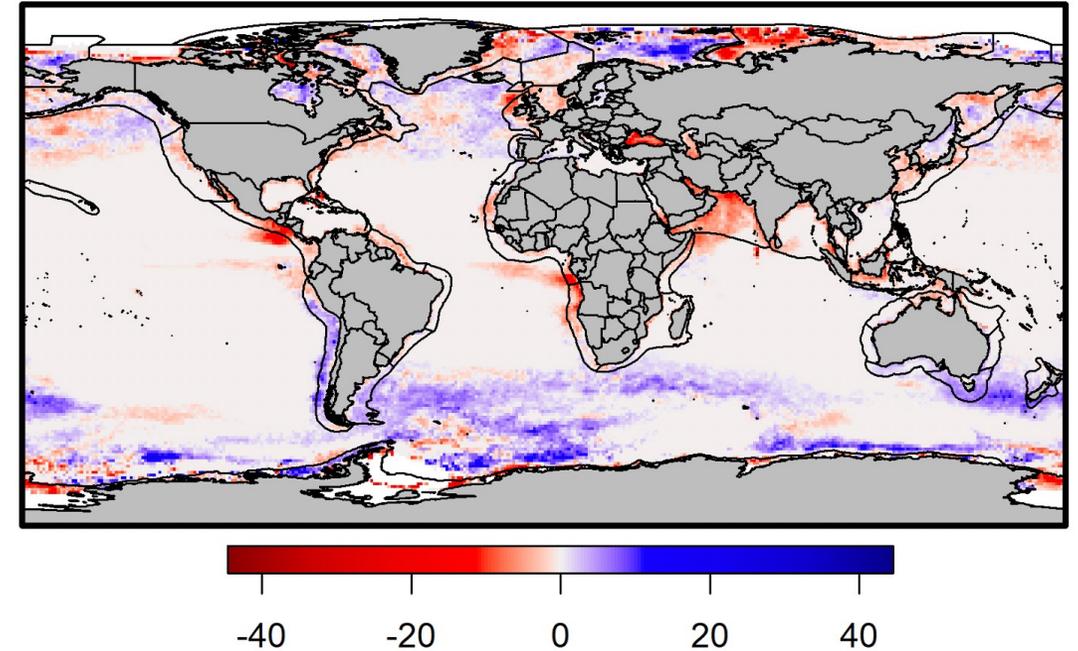
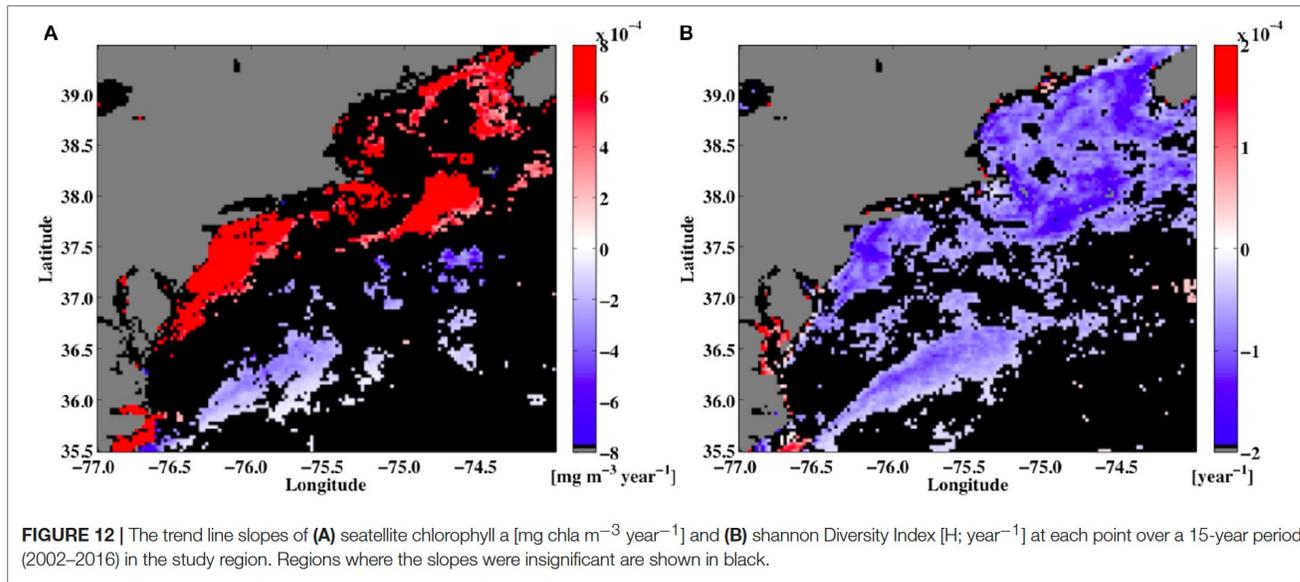
NEPAC Status

- Front-End at beta-test stage
- Working to link algorithm codes to ocean color processing software
- GPCODE has been modified:
 - Bootstrapping
 - Optimizing tree diversity through entropy.
 - Maximum Probability Regression.
- Started testing multiple satellite data sets
- Two summer interns → Prim. Prod. Archive.
- Growing interest to compare blended satellite products with observations.
- Obtaining training data sets for hyperspectral algorithms (2021 IRAD)
- Seeking funding for phytoplankton pigment and functional type alg.

Ocean Color Hyperspectral Inverse Modeling Results

Detecting trends in phytoplankton diversity and biomass^{d,e,f}

Global trend in diatoms [Theil-Sen slope % decade⁻¹]^a



- (a) 2021: Friedland, K. D., **J. R. Moisan**, A. Maureaud, D. Brady, A. Davies, S. Bograd, R. Watson, and Y. Roussouw, Trends in phytoplankton communities within large marine ecosystems diverge from the global ocean, *Canadian Journal of Fisheries and Aquatic Science*, Accepted.
- (b) 2020: Friedland, K. D., R. E. Morse, N. Shackell, J. C. Tam, J. L. Morano, **J. R. Moisan**, and D. C. Brady, Changing physical conditions and lower and upper trophic level responses on the US Northeast Shelf, *Frontiers in Marine Science*, <https://doi.org/10.3389/fmars.2020.567445>.
- (c) 2020: Wang, G., and **J. R. Moisan**, Remote Sensing of Phytoplankton Pigments, In “Plankton Communities”, Leonel Pereira (ed.), Open Access, IntechOpen, London, ISBN 978-1-83968-609-2.
- (d) 2017: Moisan, T. A., K. Rufty, **J. R. Moisan**, and M. Linkswiler. Satellite observations of phytoplankton functional type spatial distributions, phenology, diversity and ecotones, *Frontiers in Mar. Sci.*, **4**:189, doi: 10.3389/fmars.2017.00189.
- (e) 2013: Moisan, T. A. H., **J. R. Moisan**, and M. A. Linkswiler. Algorithm development for predicting biodiversity based on phytoplankton absorption, *Cont. Shelf Res.*, **55**, 17-28.
- (f) 2011: **Moisan, J. R.**, T. A. Moisan, and M. A. Linkswiler. An Inverse Modeling Approach to Estimating Phytoplankton Pigment Concentrations from Phytoplankton Absorption Spectra, *J. Geophys. Res.*, **116**, C09018, doi: 10.1029/2010JC006786.

GP Project Directions

